Transport Equations: Mathematical and Discrete Issues

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Turbulence and Modeling

- For this subject, let me quote Heisenberg.
 - "When I meet God, I am going to ask him two questions: Why relativity? And Why turbulence? I really believe he will have an answer for the first."
 - "An expert is someone who knows the worst mistakes that can be made in his subject, and how to avoid them"
- Caveat: I certainly do not consider myself an expert in either turbulence or the modeling of turbulence.
 - However, I have studied these subjects with experts. In addition, my graduate work and dissertation concerned turbulence.
 - My focus is on Mathematics and Numerical Analysis

Outline of Discussion

- Brief discussion of 3 hierarchel models
 - One-equation model
 - Two-equation model
 - Some issues in solving two-equation model
 - Full Reynolds stress model (RSM)
- Brief discussion of current solution algorithm
- Some numerical results
- Perspective on solving transport equations
 - Mathematical foundation and formulation
 - Strong solution algorithms
 - Stiffness issues
 - Positivity and realizability
 - Linear and nonlinear stability
 - Summary of some key solution algorithm properties

Compressible Reynolds-Averaged Navier-Stokes Equations

Integral form of the 3-D Reynolds-averaged Navier-Stokes equations

$$\iiint_{\mathcal{V}} \frac{\partial \mathbf{W}}{\partial t} d\mathcal{V} + \iint_{\mathcal{S}} \mathcal{F} \cdot \mathbf{n} d\mathcal{S} = 0,$$

where the solution vector

$$\mathbf{W} = \begin{bmatrix} \rho & \rho u & \rho v & \rho w & \rho E \end{bmatrix}^T$$

and the flux density tensor $\mathcal{F}=\mathcal{F}_c+\mathcal{F}_v$, with

$$\mathcal{F}_{c} = \begin{bmatrix} \rho q & \rho u q + p \mathbf{e}_{x} & \rho u q + p \mathbf{e}_{y} & \rho w q + p \mathbf{e}_{z} & \rho H q \end{bmatrix}^{T}$$
$$\mathcal{F}_{v} = -\begin{bmatrix} 0 & \overline{\tau} \cdot \mathbf{e}_{x} & \overline{\tau} \cdot \mathbf{e}_{y} & \overline{\tau} \cdot \mathbf{e}_{z} & \overline{\tau} \cdot \mathbf{q} + k \nabla T \end{bmatrix}^{T}$$

where $\mathbf{q} = u\mathbf{e}_x + v\mathbf{e}_y + w\mathbf{e}_z$, and $\overline{\tau}$ is the stress tensor.

One Equation Turbulence Model

 One-equation model of Spalart-Allmaras (La Recherche Aerospatiale 1 1994)

$$\frac{\partial \tilde{\nu}}{\partial t} + u_j \frac{\partial \tilde{\nu}}{\partial x_j} = C_{b1}(1 - f_{t2})\hat{S}\tilde{\nu} - \left[C_{w1}f_w - \frac{C_{b1}}{\kappa^2}f_{t2}\right]\left(\frac{\tilde{\nu}}{d}\right)^2 \\ + \frac{1}{\sigma}\left[\frac{\partial}{\partial x_j}\left((\nu + \tilde{\nu})\frac{\partial \tilde{\nu}}{\partial x_j}\right) + C_{b2}\frac{\partial \tilde{\nu}}{\partial x_j}\frac{\partial \tilde{\nu}}{\partial x_j}\right]$$

where the eddy viscosity is determined from

$$\mu_t = \rho \tilde{\nu} f_{v1},$$

and

$$f_{v1} = \frac{\chi^3}{x^3 + c_{v1}^3}, \quad \chi = \frac{\tilde{\nu}}{\nu}, \quad \nu = \frac{\mu}{\rho}, \quad \hat{S} = \Omega + \frac{\tilde{\nu}}{k^2 d^2} f_{v2}$$

- Implementation
 - Advection terms: First order, non-conservative discretization
 - Boundary Conditions: solid surfaces ($\tilde{\nu} = 0$); free stream ($\tilde{\nu} = 3\nu_{\infty}$)

Two Equation Turbulence Model

• A form of the two-equation k- ω model of Wilcox (Turbulence Modeling for CFD 2006) is given by

$$\frac{\partial k}{\partial t} + \tilde{u}_j \frac{\partial k}{\partial x_j} = \frac{1}{\bar{\rho}} \frac{\partial}{\partial x_j} \left[\left(\mu + \sigma^* \frac{\bar{\rho}k}{\omega} \right) \frac{\partial k}{\partial x_j} \right] + P_k - D_k$$
$$\frac{\partial \omega}{\partial t} + \tilde{u}_j \frac{\partial \omega}{\partial x_j} = \frac{1}{\bar{\rho}} \frac{\partial}{\partial x_j} \left[\left(\mu + \sigma \frac{\bar{\rho}k}{\omega} \right) \frac{\partial \omega}{\partial x_j} \right] + \frac{\sigma_d}{\omega} \frac{\partial k}{\partial x_j} \frac{\partial \omega}{\partial x_j} + P_\omega - D_\omega$$

where the production and destruction terms are given by

$$P_k = \frac{\mu_t}{\bar{\rho}} |\mathbf{\Omega}|^2, \quad P_\omega = \alpha \frac{\omega}{k} \frac{\mu_t}{\bar{\rho}} |\mathbf{\Omega}|^2, \quad D_k = \beta^* k \omega, \quad D_\omega = \beta \omega^2$$

• To reduce the magnitude of μ_t when the production of the turbulence energy exceeds the dissipation rate, a stress limiter is introduced:

$$\tilde{\omega} = \max\left\{\omega, \quad C_{lim}\sqrt{\frac{2\bar{\Omega}_{ij}\bar{\Omega}_{ij}}{\beta^*}}\right\}, \qquad C_{lim} = 0.95$$

Two Equation Turbulence Model

• Solid wall boundary conditions (Menter [AIAA J. 1994]):

$$k = 0,$$
 $\omega = \omega_w = \frac{60\mu_1}{\rho_1 \beta(d_1)^2}.$

• Free-stream boundary conditions [NASA TM 1998-208444]:

$$k = k_{\infty} = 9 \times 10^{-9} u_{\infty}^2, \quad \omega = \omega_{\infty} = 1 \times 10^{-6} \left(\frac{u_{\infty}^2}{\nu_{\infty}}\right)$$

• The turbulent viscosity is computed by

$$\mu_t = \frac{\rho k}{\tilde{\omega}}$$

Issues in Solving Two-Equation Model

- Ensuring positivity of dependent variables of turbulence model equations
- Imbalance between production and destruction terms
 - Such an imbalance occurs in regions where ω becomes small
 - \ast Allows disturbances in the strain rate
 - $\ast\,$ Results in large values of μ_t
 - * Production term becomes very large when ω goes to zero and μ_t is finite
 - To prevent excessively large ${\cal P}_k$ term, we use the constraint suggested by Menter

 $\tilde{P}_k = \min(P_k, 20D_k)$

Full Reynolds Stress Model (RSM)

• A form of the SSG/LRR full RSM [Eisfeld, 2004] is given by

$$\frac{\partial \bar{\rho} \hat{R}_{ij}}{\partial t} + \frac{\partial \bar{\rho} \hat{u}_k \hat{R}_{ij}}{\partial x_k} = \bar{\rho} P_{ij} + \bar{\rho} \Pi_{ij} - \bar{\rho} \varepsilon_{ij} + \bar{\rho} D_{ij} + \bar{\rho} M_{ij}$$
$$\frac{\partial \bar{\rho} \omega}{\partial t} + \frac{\partial \bar{\rho} \hat{u}_k \omega}{\partial x_k} = \frac{\alpha_\omega \omega}{\hat{k}} \frac{\bar{\rho} P_{kk}}{2} - \beta_\omega \bar{\rho} \omega^2 + \frac{\partial}{\partial x_k} \left[\left(\bar{\mu} + \sigma_\omega \frac{\bar{\rho} \hat{k}}{\omega} \right) \frac{\partial \omega}{\partial x_k} \right] + \sigma_d \frac{\bar{\rho}}{\omega} \max \left(\frac{\partial \hat{k}}{\partial x_j} \frac{\partial \omega}{\partial x_j}, 0 \right)$$

where the production term is given by

$$\bar{\rho}P_{ij} = -\bar{\rho}\hat{R}_{ik}\frac{\partial u_j}{\partial x_k} - \bar{\rho}\hat{R}_{jk}\frac{\partial u_i}{\partial x_k}$$

and the dissipation is modeled with ($arepsilon=C_{\mu}\hat{k}\omega$, $\hat{k}=\hat{R}_{ii}/2$)

$$\bar{\rho}\varepsilon_{ij} = \frac{2}{3}\bar{\rho}\varepsilon\delta_{ij}$$

Full Reynolds Stress Model (RSM)

• Note the following

$$\bar{\rho}\hat{R}_{ij} = -\tau_{ij} = \overline{\rho u_i'' u_j''}$$

• The diffusion term is modeled by

$$\bar{\rho}D_{ij} = \frac{\partial}{\partial x_k} \left[\left(\bar{\mu}\delta_{k\ell} + D\frac{\bar{\rho}\hat{R}_{k\ell}}{C_{\mu}\omega} \right) \frac{\partial\hat{R}_{ij}}{\partial x_{\ell}} \right]$$

- For details on modeling of the pressure-strain correlation $(\bar{\rho}\Pi_{i,j})$
 - NASA Turbulence Modeling Resource (TMR)
 - Eisfeld [DLR, ISSN 1614-7790, 2004]
- Reynolds stresses are zero at solid wall boundary, and BC for ω is the same as for Wilcox 2006 model.

RK/Implicit Scheme

• Consider a multistage scheme (e.g, Runge-Kutta)

$$\mathbf{W}^{(0)} = \mathbf{W}^{n}$$
$$\mathbf{W}^{(q)} = \mathbf{W}^{(0)} - \alpha_q \mathcal{L} \mathbf{W}^{(q-1)}, \quad q = 1, ..., s$$
$$\mathbf{W}^{n+1} = \mathbf{W}^{(q)}$$

 $\ensuremath{\mathcal{L}}$ is the complete difference operator.

• Suppose we introduce an implicit preconditioner \mathcal{P}^{-1} on stage q

$$\mathbf{W}^{(q)} = \mathbf{W}^{(0)} - \alpha_q \, \mathcal{P}_q^{-1} \, \mathcal{L} \, \mathbf{W}^{(q-1)}$$

where

$$\mathcal{P}_{q} = \left(\frac{1}{CFL\left(\Delta t\right)}\right)I + \left(\frac{\partial \mathbf{R}}{\partial \mathbf{W}}\right)_{q}$$

- The resulting scheme \in of the general class of implicit RK schemes
- Now, we have to determine the discrete implicit operator.

Inverse of Implicit Operator

• Consider the preconditioner

$$\mathcal{P}_{q} = \left(\frac{1}{CFL\left(\Delta t\right)}\right)I + \left(\frac{\partial \mathbf{R}}{\partial \mathbf{W}}\right)_{q}$$

- The inverse of the preconditioner \mathcal{P}_q is approximated.
- Stiff discrete equations are considered.
 - Point SGS and partial line SGS relaxation used to approx. \mathcal{P}_q^{-1} .
 - Sufficient approx. with 2 SGS sweeps for each relaxation type
 - CFL is between $10^3 \ {\rm and} \ 10^6$
- Implementation details of RK/Implicit scheme given in
 - Rossow [JCP 2007]
 - Swanson et al. [JCP 2007],
 - Swanson, Rossow [Comput. Fluids 2011]

Numerical Results

- Numerical solutions for airfoil flows
 - Nakayama A-Airfoil
 - RAE Airfoil
 - NACA 4412 Airfoil

Convergence Histories for Nakayama A-Airfoil

k- ω Model, $\alpha = 0^{o}$, M = 0.088, $Re = 1.2 \times 10^{6}$



Convergence Histories for Nakayama A-Airfoil

k- ω Model, $\alpha = 0^{o}$, M = 0.088, $Re = 1.2 \times 10^{6}$



Results for Nakayama A-Airfoil

 $k\text{-}\omega$ Model, $\alpha=0^o$, M=0.088, $Re=1.2\times10^6$



Surface Pressure and Skin-Friction Distributions (RK/Implicit Scheme)

Convergence Histories with RSM for Nakayama A-Airfoil

Reynolds Stress Model, $\alpha = 0^{o}$, M = 0.088, $Re = 1.2 \times 10^{6}$



Computations with Implicit AF Scheme and Reynolds Stress Model

Convergence Histories for RAE 2822 Airfoil: Subsonic

 $\alpha = 1.93^{o}$, M = 0.676, $Re = 5.7 \times 10^{6}$



Case 1: RK/Implicit scheme, k- ω model

Convergence Histories for RAE 2822 Airfoil: Transonic

 $\alpha = 2.79^{o}$, M = 0.730, $Re = 6.5 \times 10^{6}$



Case 9: RK/Implicit scheme, k- ω model

Results for RAE 2822 Airfoil: Transonic

 $\alpha = 2.79^{o}$, M = 0.730, $Re = 6.5 \times 10^{6}$



Computations with RK/Implicit Scheme and k- ω model

Convergence Histories for RAE 2822 Airfoil: Transonic

 $\alpha = 2.81^{o}$, M = 0.750, $Re = 6.2 \times 10^{6}$



Case 10: RK/Implicit scheme, k- ω model

Results for RAE 2822 Airfoil: Transonic

 $\alpha = 2.81^{o}$, M = 0.750, $Re = 6.2 \times 10^{6}$



Computations with RK/Implicit Scheme and k- ω model

Convergence Histories for NACA 4412 Airfoil High Angle of Attack (α)

 $\alpha = 13.87^o$, M = 0.09 , $Re = 1.52 \times 10^6$



NACA 4412: RK/Implicit scheme, k- ω model

Results for NACA 4412 Airfoil

High Angle of Attack (α)

 $\alpha = 13.87^o$, M = 0.09 , $Re = 1.52 \times 10^6$



Computations with RK/Implicit Scheme and k- ω model

Turbulent Viscosity Contours High Angle of Attack (α) $\alpha = 13.87^{\circ}$, M = 0.09, $Re = 1.52 \times 10^{6}$ Upper LE Region, 1988 Wilcox Model



Turbulent Viscosity Contours and Streamines High Angle of Attack (α) $\alpha = 13.87^{\circ}$, M = 0.09, $Re = 1.52 \times 10^{6}$ Lower LE Region, 1988 Wilcox Model



Turbulent Viscosity Contours and Streamines High Angle of Attack (α) $\alpha = 13.87^{o}$, M = 0.09, $Re = 1.52 \times 10^{6}$

Lower and Upper LE Region, 2006 Wilcox Model



Navier-Stokes (N-S) Equations

- Mathematical foundation
 - Need to solve: Problem of well-posedness, regularity, and global existence
 - Progress made for particular cases.
 - Problem in general remains unsolved
- Transport equations for turbulence models
 - Form is similar to that for N-S equations.
 - Mathematical foundation is not established.
- Thus, we do not know a priori for the transport equations
 - If a solution exist
 - Possibly, there are multiple solutions
- Existence of a realistic solution to the transport equations depends only on **numerical demonstrations**.

Mathematical Formulation for Turbulent Flow Problems

- Issues
 - Often, complete descriptions of continuous and discrete problems are not given
 - BCs are not always stated, and often are only approximate
 - Sometimes, a boundary-value problem is not even be solved for each iteration (e.g., wall BC for ω equation).
- There is usually neglect of a proper study of the mathematical properties of transport equations
 - This is even true when mean flow and turbulence model equations are weakly coupled.
 - Exception has been with the Spalart-Allmaras (SA) model.
- Even without mathematical questions, additional numerical difficulties can occur
 - When considering two or more transport equations (i.e., PDE related to length scale equation used)
 - Increased stiffness of system of equations

Mathematical Formulation for Turbulent Flow Problems

• Thus, the reliability of using such models on a routine basis for industrial application can be problematic.

Conjecture 1 The construction of a reliable and efficient algorithm for solving the transport equations of a turbulence model, either weakly or strongly coupled to the Reynolds average N-S (RANS) equations, is highly unlikely, in general, without complete mathematical and numerical analysis of the system or systems of equations.

Another View of Turbulence Modeling

- Generally, we consider turbulence modeling a direct problem
 - Model parameters determined by comparing solutions and data for range of problems.
 - Solve RANS equations with given turbulence model
- One can consider turbulence modeling an inverse problem, as suggested by S. Langer [Habilitation, 2017].
 - Then, we have a parameter identification problem
 - Reconstruct the eddy viscosity from given data
- Indicators that inverse problem is ill-posed
 - More than one turbulence model of a given type gives essentially same surface C_p and C_f values.
 - Small perturbation in C_p variation can yield different solution and μ_t .
- Such observations suggest that turbulence modeling by its mathematical nature is an ill-posed problem.

Developing of Turbulence Models

- Developers of turbulence models
 - Again, generally give little attention to solving equations of model
 - Difficulties arise due to both stiffness issues and source terms
- Need to consider two factors in modeling of turbulence
 - Ultimately, the success of model depends on heuristics of model.
 - Models with more than one equation are not based upon first principles derivations.
- Consider a two-equation model
 - One equation is turbulence kinetic energy (TKE) equation derived from first princples.
 - Second equation is used to give a length scale; it is kludged up (as Peter Bradshaw might say)
 - * Form of equation based on that of TKE equation
 - * Example: equation for turbulence dissipation rate (ω)
 - * Eddy viscosity is $\mu_t = \rho k / \omega$.

Developing of Turbulence Models

- The two factors previously given suggest that there is flexibility in changing model to make amenable to solving equations.
- The essential requirements for changing model
 - Must maintain integrity of model.
 - Must verify the modified model for standard test cases used to calibrate parameters of model.

Strong Solution Algorithms

- Assume RANS and transport equations are weakly coupled.
- Numerical algorithms for both RANS and transport equations
 - Should both belong to the class of strong solution algorithms
 - Would like algorithms to be the same to allow eventually a strongly coupled solver
 - Must be highly implicit: L-stability is desired to solve stiff systems.
 - Algorithm must be unconditionally stable in linear sense, including BCs.
 - Fluid dynamics equations highly nonlinear in general \rightarrow desirable to have nonlinear stability
 - Same stability properties should apply to solvers for mean flow and transport equation.

Stiffness Issues Concerning Transport Equations

- No generally accepted mathematical definition of stiffness.
- Some examples of stiff systems:
 - Linear constant coefficient system with large stiffness ratio
 - Stability requirements rather than accuracy restrict stepsize
- Two fundamental sources of stiffness
 - Systems of PDEs themselves
 - * Large condition number (stiffness number) for compressible Euler or Navier-Stokes equations as $M\to 0$
 - * Source terms of turbulence modeling equations create stiff systems.
 - Discrete resolution of viscous terms using stretched grids results in stability determining stepsize.
- Stiff ODEs require strongly implicit schemes for an effective solution algorithm.

Positivity and Realizability

- Need for Positivity Preserving Scheme
 - Requirement of positivity for thermodynamic variables of mean flow equations
 - For two-equation model (e.g., $k \omega$ model of Wilcox),
 - $\ast\,$ Need to ensure positivity of k and $\omega\,$
 - * Eddy viscosity $\mu_t = \rho k/\omega > 0$
 - For full RSM, dependent variables include Reynolds stresses
 - $\ast\,$ Need to ensure positivity of normal Reynolds stresses and $\omega\,$
 - * Realizability conditions: $\overline{\rho u_i'' u_i''} \ge 0$, $(\overline{\rho u_i'' u_j''})^2 \le \overline{\rho u_i'' u_i''} \overline{\rho u_j'' u_j''}$
- For turbulence models, there are 3 methods used to ensure positivity of dependent variables.
 - Explicitly enforce positivity
 - Replace k, ω with $\ln(k)$, $\ln(\omega)$ (Ilinca and Pelletier, [IJTS 1999])
 - $\ast\,$ Need wall function since k vanishes at solid wall boundary
 - * Approach can experience difficulties with upper bound on μ_t .

Positivity and Realizability

- Replace ω with $\ln(\omega)$ (Bassi et al, [Comput Fluids, 2005])
 - Limiting k
 - Imposing lower bound on ω using realizability conditions
 - * Positive normal Reynolds stresses
 - * Satisfying Schwarz inequality for Reynolds shear stresses.
- Construct M-matrix (Mor-Yossef [JCP 2006])
 - A semi-positive matrix with positive eigenvalues
 - Used to approximate Jacobian of transport equations
 - A non-singular M-matrix has convergent regular splitting and positive inverse.
 - These properties guarantee
 - * Unconditional linear stability
 - * Positivity of dependent variables in turbulence equations
 - Note: M-matrix approach does not appear to have a counterpart for mean flow equations.

Linear and Nonlinear Stability

- Linear stability
 - Local mode analysis (Fourier transform of linear equations)
 - For initial-boundary value (IBV) problems, it is beneficial to consider energy stability.
 - * Allows variable coefficents
 - * Includes boundary conditions
- Eigensystem analysis for IBV problems
- With Krylov method, using Arnoldi algorithms, one can determine stability behavior of actual discrete problem being solved (Langer [JCP 2014]).

Linear and Nonlinear Stability

- Nonlinear stability
 - Generally not included in constructing numerical algorithms
 - Nonlinearity can have significant effects such as producing spurious oscillations, occurring in
 - * Under-resolved regions
 - * Neighborhood of discontinuities such as shock waves
- For this purpose, we consider entropy stability.
- Entropy stability guarantees that the thermodynamic entropy is bounded for all time in L_2 under two conditions (Carpenter et al. [AIAA 2013]).
 - ho, T > 0
 - Boundary data results in a well-posed problem and preserves entropy estimate.
- To construct entropy stable scheme, a convex function is required.

Linear and Nonlinear Stability

- Entropy function
 - Mathematical function
 - Negative of thermodynamic entropy of gas dynamics
- Global conservation of entropy is derived from transforming (contracting)
 N-S equations using
 - Entropy function and
 - Integrating over the domain
- By mimicking each term of this equation discretely, we can obtain a semi-discrete entropy estimate.
- Using procedure of Tadmor [Acta Numerica, 2003] called comparison of arguments, conditions that guarantee entropy stability can be established.
- Considering entropy stability for algorithm to solve transport equations of turbulence model requires
 - Appropriate entropy function
 - Again, proving well-posedness for discrete problems

Summary: Solution Algorithms

• Basic elements

- Implicit scheme
 - * Required for stiff equations (L-stability)
 - * Point and line relaxation to invert implicit operator
 - * Parallelizability
- Multigrid or emulator
 - * Scheme with good smoothing properties
 - * Multistage framework for scheme to enhance eigenvalue clustering

• A class of strong algorithms

- Basic elements
- Minimized inconsistencies: implicit operator and residual function
- Newton-Krylov methods
 - * Effective preconditioner
 - * Multigrid in linear solver

Summary: Solution Algorithms

- Enhanced robustness with Adaptive Algorithms
 - Local measures of stability and resolution
 - Enriched local resolution
 - Variable order of discretization
 - Additional local relaxation due to high residuals
 - Modification of algorithm parameters
 - Modified local relaxation strategy
- Ensure positivity
- Contruct schemes with entropy (nonlinear) stability