Some Recent Developments in Turbulence Closure Modeling

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Abstract

Turbulence closure models are central to a good deal of applied computational fluid dynamical analysis. Closure modeling endures as a productive area of research. This review covers recent developments in elliptic relaxation and elliptic blending models, unified rotation and curvature corrections, transition prediction, hybrid simulation, and data-driven methods. The focus is on closure models in which transport equations are solved for scalar variables, such as the turbulent kinetic energy, a timescale, or a measure of anisotropy. Algebraic constitutive representations are reviewed for their role in relating scalar closures to the Reynolds stress tensor. Seamless and nonzonal methods, which invoke a single closure model, are reviewed, especially detached eddy simulation (DES) and adaptive DES. Other topics surveyed include data-driven modeling and intermittency and laminar fluctuation models for transition prediction. The review concludes with an outlook.

Keywords

turbulent flow, laminar to turbulent transition, hybrid simulation, elliptic blending, data-driven modeling
1. INTRODUCTION

Turbulence closure modeling has a storied, if perhaps inelegant, history. Luminaries including G.I. Taylor, Ludwig Prandtl, A.N. Kolmogoroff, and Theodore von Kármán, among others, invoked analogies to kinetic theory and Brownian movement, as well as concepts of universality and dimensional scaling, to lay foundations of the subject. Perhaps prophetically, representing turbulent transport by an eddy viscosity underlies most of this seminal work. Today, the framework of closure modeling remains the same as that of the early pioneers: Turbulent transport is modeled by eddy viscosities, velocity covariances, and correlation timescales.

From its outset, closure modeling has been a mixture of analysis, empiricism, and analogy, directed to rather practical aims. After a period of time, during which the goal of most practical turbulence modeling was predicting bulk parameters, computer power advanced to the state that solving partial differential equations for turbulent transport became practical. Computer power led to the modern era of turbulence modeling, which is sometimes said to have begun after the 1968 Stanford conference on turbulent boundary layers. At the outset of the 1970s, researchers at Imperial College led by Brian Spalding and Brian Launder took a prominent role, with many notable contributors including Kemo Hanjalic, Wolfgang Rodi, Peter Bradshaw, Mike Gibson, Bill Jones, and Steve Pope, among others. The $k - \varepsilon$, $k - \omega$, and various Reynolds stress transport models date from this time period, albeit with precedents from earlier times and with many contributions from outside of the Imperial College circle. Algorithms for Reynolds-averaged Navier–Stokes (RANS) computation, as well as the idea of the commercial computational fluid dynamics (CFD) code, also stem from this period. Algorithms have developed systematically and progressively, and commercial CFD has grown explosively and profitably. Closure models continue to play a major role in applied CFD and remain an area of active research and development.

At times it has been suggested that RANS models would wither away and be replaced by large-eddy simulation (LES) as a consequence of rapidly increasing computer power. Certainly, greater reliance on the unaveraged Navier–Stokes equations means more accurate results. But the demise of RANS closures has not happened: The expense in computer time, man-hours, and requirements for high-resolution gridding continue to make extensive use of LES unfeasible. Laurence (2002) provided the insight that advances in computer power have enabled RANS analysis to be applied more broadly to increasingly complex configurations and that, rather than withering away, its use has expanded. This insight has been echoed by Hanjalić (2005), who noted that RANS analysis has become ingrained in engineering practice through increased reliance on user-friendly commercial CFD codes. Closure models will remain a critical element of computational fluid dynamics analysis into the foreseeable future (Slotnick et al. 2014).

Despite the persistence of RANS models in applied CFD, the niche for eddy-resolving simulation is also growing. Most relevant to the present review is the marriage of closure models and eddy-resolving simulation, termed hybrid methods (Frohlich & Von Terzi 2008). In that vein, so-called seamless methods, which invoke the closure model throughout the whole computational domain, are within the scope of this review.

Turbulence closure modeling has been the subject of many reviews (e.g., Launder 1989, Speziale 1991, Gatski & Rumsey 2002, Hanjalić & Kenjereš 2008) and texts (e.g., Wilcox 1998, Pope 2001, Durbin & Pettersson-Reif 2010, Leschziner 2016) and is commonly reviewed in books on CFD and manuals on CFD software. The subject of turbulence closure modeling is broad, and research continues on many fronts. This review surveys a few selected areas in which closure modeling has advanced in recent years.
2. SCALAR AND TENSOR CLOSURE MODELS

When the Navier–Stokes equations are averaged, they become unclosed. In other words, there are more variables than equations. The mean flow equation contains the velocity covariance tensor, also known as the Reynolds stresses. The equations for the covariance tensor contain both higher-order correlations and nonlocality via the pressure–velocity correlation (see Pope 2001). Closure models are empirical formulas and equations that increase the number of equations to equal the number of unknowns. To meet the needs of modern CFD, closure models invoke transport equations, which provide fields of turbulence variables. Sometimes, the variables are particular to the model, having no unambiguous physical meaning; always, the aggregate model contains ambiguous features and relies heavily on empiricism.

Transport equations used in closure modeling can be grouped into those with scalar dependent variables and those with tensor variables. Examples of the former are equations for $k$, $\varepsilon$, $\omega$, and $\tilde{\nu}$, as in the $k - \varepsilon$, $k - \omega$, or Spalart–Allmaras (S-A) models. The latter are Reynolds stress transport models, with $\overline{\nu u_i}$ or $\overline{\nu u_i}$, for example, as dependent variables. Well-established models are not reviewed here. Indeed, only two topics, elliptic blending and algebraic stress constitutive relations, are discussed.

2.1. Elliptic Blending Models

Elliptic blending and elliptic relaxation are motivated by anisotropy caused by wall proximity. As the surface is approached, the wall-normal component of intensity decreases relative to the other two. Scalar, elliptic models have two components: The first, elliptic component incorporates wall influence as a function of the turbulent length scale and of the geometry, and the second component introduces a scalar measure of anisotropy.

The second component might seem contradictory because scalars are isotropic. The measure of anisotropy is characterized as the ratio of wall-normal to total velocity variance, $v^2/k$. This is abstracted as a scalar. It is denoted either as $\zeta$ (Hanjalić et al. 2004) or as $\phi$ (Laurence et al. 2004). Its isotropic value is 2/3. The eddy viscosity $\nu_T$ is defined as either $C_\mu \zeta kT$ or $C_\mu \phi kT$, where $T$ is a correlation timescale.

The elliptic equation for $f$ is

$$L^2 \nabla^2 f - f = -R,$$

where $R$ is a simplified pressure–strain formula in elliptic relaxation and $R = 1$ in elliptic blending. Both the $\zeta - f$ and $\phi - f$ models invoke return to isotropy and rapid pressure–strain closures for $R$. As an example, in the $\zeta - f$ model (Hanjalić & Kenjereš 2008),

$$R = \frac{1}{T} \left( \frac{2}{3} - \zeta \right) \left( C_1 + C_2 \frac{P}{\varepsilon} \right),$$

where $C_1$ and $C_2$ are constants, $T$ is the turbulence timescale, and $P/\varepsilon$ is the ratio of production to dissipation of turbulent kinetic energy. $L$ is a turbulence length scale—some combination of the integral $k^{1/2}/\varepsilon$ and the Kolmogoroff $(\nu^3/\varepsilon)^{1/4}$ scales.

The derivation of Equation 1 relates it to a two-point correlation function that appears in an exact formula for pressure–strain covariance in homogeneous turbulence (Durbin 1993, Manceau 2015). $L$ thereby arises as a correlation length scale (Manceau et al. 2001). However, in practice, Equation 1 is used to represent nonhomogeneous wall effects.

In the $\zeta - f$ and $\phi - f$ models, Equation 1 is combined with a transport equation for the anisotropy measure, $\zeta$ or $\phi$. The transport equation for either $\zeta$ or $\phi$ contains $f$ as a source term.
For example, using $\zeta$ as the dependent variable, the equation is

$$\frac{D\zeta}{Dt} = f - \frac{P_k}{k} \zeta + \nabla \cdot \left[ \left( \nu + \frac{1}{\sigma_c} \right) \nabla \zeta \right] + X,$$

where $X$ represents additional terms that appear in some versions of this formulation. Hence, in effect, the wall modifies the pressure–strain term via the $f$ equation.

Billard & Laurence (2012) developed an elliptic-blending variant termed $EB-\nu^2/k$. The blending function satisfies

$$L^2 \nabla^2 \alpha - \alpha = -1,$$

with $\alpha = 0$ on walls and $\alpha \rightarrow 1$ far from the surface. In Equation 3, $f$ is replaced by

$$(1 - \alpha^2) f_w + \alpha^2 f_h$$

to interpolate between homogeneous $f_h$ and near-wall $f_w$ pressure–strain formulations, with the cubic power chosen to give a correct near-wall asymptote [expressions for $f_h$ and $f_w$ can be found in Billard & Laurence (2012) and Lardeau & Billard (2016)]. This formulation is more numerically robust than the elliptic-relaxation version. It predicts separated flows more accurately (Billard et al. 2012).

An extension of this formulation was developed by Lardeau & Billard (2016), with extra production terms attributed to the lag between stress and strain rate, which occurs naturally in Reynolds stress transport models. Lardeau & Billard (2016) developed a scalar model of the form of Equation 3 with a contribution to the last term $X$ that was derived from an analogy to Reynolds stress transport. Predictions of the lag elliptic blending ($lag~EB$) $k - \epsilon$ model improve upon the original model, as is illustrated by the example in Figure 1. Accurate prediction of separation and reattachment is one of the holy grails of closure modeling, and the $EB-\nu^2/k$ and $lag~EB$ $k - \epsilon$ models appear to be significant advancements toward this goal. The $lag~EB$ model has the advantage that it depends on the rate of rotation tensor $\Omega_{ij}$ so that it can be adapted to rotation and curvature by the method of Equation 17 in Section 2.3.

Elliptic blending originated as a method for tensoral closure modeling (Manceau & Hanjalić 2002, Manceau 2005). In that case, the pressure–strain (more correctly, redistribution) closure is written as

$$\phi_{ij} = (1 - \alpha^2) \phi^w_{ij} + \alpha^2 \phi^h_{ij},$$

where $\alpha$ solves the elliptic Equation 4. The wall term depends on a normal direction, which is defined by $\hat{n} = \nabla \alpha / |\nabla \alpha|$. An explicit formula is

$$\phi^w_{ij} = -\frac{5}{T} \left[ \frac{\hat{n}_i \nabla_{j} \hat{n}_i + \hat{n}_j \nabla_{i} \hat{n}_i}{\nabla \cdot \nabla \hat{n}_i \hat{n}_i \hat{n}_j} + \frac{1}{2} \frac{\nabla \cdot \nabla \hat{n}_i \hat{n}_j}{\nabla \cdot \nabla \hat{n}_i \hat{n}_i} \hat{n}_i \hat{n}_i \hat{n}_j + \delta_{ij} \right].$$

The homogeneous term $\phi^h$ can be any redistribution model, such as the Speziale–Sarkar–Gatski formulation (Speziale et al. 1991). $\phi_{ij}$ appears in the Reynolds stress transport equation, written symbolically as

$$\frac{D\Pi_{ij}}{Dt} = P_{ij} + \phi_{ij} - \frac{2}{3} \epsilon \delta_{ij} + \delta_{ij} T_{ij},$$

where $P$ is production, $T$ is transport, and anisotropic dissipation is lumped into the redistribution term. The wall term is necessary to correct the wrong anisotropy built into homogeneous redistribution models. The wall term, termed pressure echo, has been recognized since the advent of Reynolds stress transport closure modeling (Durbin & Pettersson-Reif 2010). The pressure echo is a term in an elliptic, Green function solution to Poisson’s equation. Manceau (2015) reviewed how that exact but unclosed term leads to elliptic blending. Billard et al. (2012) presented applications of an elliptic blending Reynolds stress model to flow separation with some impressive results,
but also with the warning that “more physical RSM’s do capture some features of the flows better than the simpler linear EVM’s, although the improvements are not always consistent or uniform across different cases” (p. 146).

Elliptic relaxation and elliptic blending models have also been applied to heat transfer and to environmental flows (Kenjereš et al. 2005; Choi & Kim 2006, 2008; Hanjalić & Kenjereš 2008; Shin et al. 2008; Dehoux et al. 2011, 2017). Improved predictions are obtained for natural and mixed convective heat transfer.
2.2. Algebraic Stress Constitutive Models

A scalar closure must nevertheless result in a representation for the Reynolds stress second-order tensor. The connection is provided by a constitutive formula. The simplest formula is the linear, stress–strain rate, eddy viscosity relation.

The notion of eddy viscosity dates to Boussinesq in the nineteenth century. A systematic development of algebraic constitutive relations from Reynolds stress transport models began in the 1970s, with Rodi’s (1976) algebraic stress approximation and Pope’s (1975) closed form solution to Rodi’s approximation via an integrity basis expansion (see Wallin & Johansson 2000). Over the years, researchers have enlarged upon that early work. The subjects of algebraic stress models (ASMs) and explicit algebraic stress models (EASMs) are now well established (Gatski & Jongen 2000). Those topics are mentioned in the present review because of renewed interest in using algebraic constitutive models in conjunction with recent developments in scalar closure and hybrid modeling.

The general notion is that mean flow gradients provide a tensoral representation that can be used to connect scalar turbulence variables to Reynolds stresses:

\[-\overline{u_i u_j} = kF_{ij}(T \partial_k U_\ell) - \frac{2}{3} k \delta_{ij}.\]

Here the scalar model provides the turbulent kinetic energy \(k\) and timescale \(T\). The timescale might be \(1/\omega\) or \(k/\varepsilon\). Eddy viscosity transport models do not have a timescale equation. In that case, Spalart (2000) replaces \(T\) by \(1/|\nabla U| = (\partial_j U_i - \partial_i U_j)^{-1/2}\).

The functional dependence of Equation 9 becomes an algebraic stress model upon invoking representation theorems for tensor functions of tensors, \(F_{ij}(\partial_k U_\ell)\). To that end, the velocity gradient is replaced by its symmetric and antisymmetric components,

\[S^{ij} = \frac{1}{2}(\partial_j U_i + \partial_i U_j)T,\]

\[\Omega^{ij} = \frac{1}{2}(\partial_j U_i - \partial_i U_j)T,\]

in nondimensional form. If \(F_{ij}\) is expanded in powers of \(S\) and \(\Omega\), then representation theorems show that the expansion contains 10 independent terms, with coefficients that depend on scalar invariants (Gatski & Jongen 2000, Gatski & Rumsey 2002).

In two dimensions, the general form of Equation 9 reduces to

\[-\overline{uu} + \frac{2}{3} k I = C_\mu k S^2 + C_1 k (S^2 : \Omega^2 - \Omega^2 : S^2).\]

The linear eddy viscosity model is \(C_1 = 0\). The coefficients \(C_\mu\) and \(C_1\) are functions of the invariants \(tr(S^2)\) and \(tr(\Omega^2)\). Other dependencies, such as on the turbulent Reynolds number \(v_T/\nu\) or on the wall distance \(\sqrt{k d}/\nu\), may also be introduced. For a two-dimensional (2D) mean velocity field in three dimensions, an additional term, \(C_2 k [S^2 - \frac{1}{3} tr(S^2)]I\), is added. In parallel shear flow, the \(C_\mu\) term in Equation 11 contributes a shear stress; the nonlinear terms contribute to normal stress.

As an ASM, the coefficients in Equation 11 are arbitrary functions of the invariants. They might be specified empirically by fitting data. If they are found by solving an equilibrium approximation to a Reynolds stress transport model (see the sidebar titled The Algebraic Stress Approximation), it becomes an EASM. Then the coefficients are determined by the Reynolds stress closure.

Although an explicit solution can be found for the general, quasi-linear redistribution closure (Durbin & Pettersson-Reif 2010) via integrity basis expansions, it is too highly nonlinear to be of value. In general, nonlinear constitutive models introduce numerical stiffness. Taulbee (1992) and Wallin & Johansson (2000) developed a promising truncation. By selecting particular coefficients...
THE ALGEBRAIC STRESS APPROXIMATION

One way to devise a constitutive equation is to reduce a Reynolds stress closure model to a set of algebraic equations. In homogeneous turbulence, the Reynolds stress transport equation is of the form

\[ \frac{d\overline{\tau}_{ij}}{dt} = P_{ij} + \phi_{ij} - \frac{2}{3} \varepsilon \delta_{ij}, \]

where \( P_{ij} \) is the rate of production of Reynolds stress and \( \phi_{ij} \) is the redistribution between components of the stress tensor. Redistribution incorporates anisotropy of dissipation, so only the isotropic part \( \varepsilon \delta_{ij} \) appears. One half of the trace of this equation is the turbulent kinetic energy equation

\[ \frac{d}{dt} k = P - \varepsilon. \]

The equilibrium stress approximation is \( d (\overline{\tau}_{ij} / k) = 0 \). Substituting from above, this gives

\[ P_{ij} + \phi_{ij} - \frac{2}{3} \varepsilon \delta_{ij} - (P - \varepsilon) \overline{\tau}_{ij} / k = 0. \]

This is a system of equations for \( \overline{\tau}_{ij} \) as an implicit function of the mean velocity gradient \( \partial_i U_j \).

in the general linear redistribution model, Taulbee (1992) obtained the equilibrium algebraic equation

\[ \left( C_R - 1 + \frac{P}{\varepsilon} \right) b = -\frac{8}{15} S + \frac{4}{9} (b \cdot \Omega - \Omega \cdot b) \]

for the anisotropy tensor, \( b \equiv \overline{\tau} / k - \frac{2}{3} \mathbf{I} \), where \( C_R \) is an empirical constant. The explicit solution of Equation 12 reduces to the two terms of Equation 11 in two dimensions, and to a sum of five terms of the integrity basis in three dimensions. The five-term solution is

\[ -\overline{\tau} + \frac{2}{3} \mathbf{I} = C_\mu k S^* + C_1 k (S^* \cdot \Omega^* - S^* \cdot \Omega^* + C_2 k [\Omega^* - \frac{1}{2} tr(\Omega^*) \mathbf{I}]
\]

\[ + C_3 k [S^* \cdot \Omega^* + \Omega^* \cdot S^* - \frac{2}{3} tr(\Omega^*) \cdot S^* \mathbf{I})]
\]

\[ + C_4 k (\Omega^* \cdot S^* \cdot \Omega^* - \Omega^* \cdot S^* \cdot \Omega^*). \]

This is an exact solution to the system of algebraic equations represented in Equation 12, obtained as an equilibrium approximation to a Reynolds stress model—i.e., it is an EASM. The functional forms of the coefficients are given in Wallin & Johansson (2000) and in Franke et al. (2005). They invoke the consistency condition that production equals the double inner product of stress given by Equation 14 and rate of strain, \( P = -\overline{\tau} : \mathbf{S} \). This precludes the singularity that occurred in Gatski & Speziale (1993) because production was not evaluated consistently.

There is an intuitive sense in which a five-term approximation is plausible. Because the trace of Equation 9 is an identity, it requires five components of \( \overline{\tau}_{ij} \) to be determined by the constitutive relation. That argues for at least a five-component basis. Were \( \overline{\tau}_{ij} \) known, one could solve for the five coefficients of the expansion (Gatski & Jongen 2000). When used as a model, the coefficients are given by formulas and \( \overline{\tau}_{ij} \) is predicted. The reason that the general integrity basis has more than five components is that elements of any five-component basis can vanish for certain
Rotation reduces turbulent intensity when the shear is in the direction of rotation and increases it where they are opposite. In plane Poiseuille channel flow, if the flow is left to right, convection will rotate a small material line near the lower (upper) wall in the clockwise (counterclockwise) direction. If the channel is rotated clockwise, turbulent intensity decreases near the lower wall and increases near the upper. The mean velocity profile becomes asymmetric, unlike rotating laminar channel flow.

Combinations of rate of rotation and rate of strain. So the ten-component basis provides complete generality.

The full formulation of Wallin & Johansson (2000) adds blending functions (not elliptic) to adapt it to near-wall behavior. Franke et al. (2005) tested it in aerodynamic flows, using the \( k-\omega \) model for the scalar terms. They found that the nonlinear constitutive model provided some improvement over the linear eddy viscosity model. Weinmann & Sandberg (2009) also found that either a 2D or 3D nonlinear constitutive relation provided improvement in a wing–body junction flow, while providing no benefit in separated flow over an axisymmetric hump.

Rumsey et al. (2000) made a suggestion that might improve the accuracy of algebraic stress models in curved flow. If the algebraic stress approximation is applied in a rotating frame, Coriolis acceleration \( (b \cdot \Omega^F - \Omega^F \cdot b) \) is added to the left side of Equation 12. Rumsey et al. (2000) found that replacing frame rotation by streamline curvature, by the method of Equation 17 in the next section, improved predictions of curved channel flow.

2.3. Rotation, Curvature, and Buoyancy

By their nature, scalar models do not respond appropriately to directional forces, such as those due to system rotation, streamline curvature, or buoyancy. These affect individual components of the Reynolds stress tensor. For instance, rotation can suppress or enhance turbulence, as a consequence of centrifugal acceleration in the radial direction, outward from the center of curvature (see the sidebar titled Rotating Turbulent Shear Flow). A scalar formula does not distinguish directional components of Reynolds stress, so the correct phenomenology is inherently absent.

2.3.1. Unification of rotation and curvature. However, models based on transport equations for scalar variables are workhorses of applied CFD. In the interest of improving their robustness, patches have been devised to sensitize such models to curvature and rotation. Early ideas (reviewed in Durbin 2011) were formulated for nearly parallel flows. Recent interest has been in formulations for general flows. Proposals have been made to base them on streamline curvature, but streamlines are not Galilean invariant. For that reason, efforts to literally represent streamline curvature have been supplanted by a method that unifies rotation and curvature via the rate of rotation of the principal axes of the rate-of-strain tensor (Spalart & Shur 1997, Wallin & Johansson 2002).

The rate-of-strain tensor is symmetric, so it can be expressed as

\[
S = \sum_{\alpha=1}^{3} \lambda_{\alpha} e^{\alpha} e^{\alpha},
\]

where the \( e^{\alpha} \)’s are unit eigenvectors and the \( \lambda_{\alpha} \)’s are eigenvalues. Unit vectors can only change by rotating: \( d t e^{\alpha} = \Omega_{\alpha \beta}^{S} \cdot e^{\beta} \). The rotation rate tensor \( \Omega_{\alpha \beta}^{S} \) can be computed from the mean velocity.
field. It has been used to parameterize effects of streamline curvature and system rotation, as explained below.

To begin, I provide some remarks on computing $\Omega^S$. On the grounds of computational efficiency, Spalart & Shur (1997) introduced the Spalart–Shur tensor:

$$W^{SS} \equiv S \cdot D_S S - D_S \cdot S$$

In two dimensions, Equation 15 equals the exact rotation tensor. Wallin & Johansson (2002) derived the connection of the Spalart–Shur tensor to rotation of the principal axes in three dimensions. If rotation vectors are introduced via $w_{SS}^i \equiv \frac{1}{2} \epsilon_{ijk} W^{SS}_{jk}$ for the Spalart–Shur tensor and $\Omega^S_i = \epsilon_{ijk} \omega^S_j$ for the rotation of eigenvectors, then the exact relation between the Spalart–Shur tensor and the rotation rate of the principal axes is

$$w_{SS}^i = \omega^S_i - \frac{3S^2_{ij}}{2|S|^2} \omega^S_j.$$ 

Given $w_S^{SS}$, the matrix on the right of Equation 16 can be inverted via the Cayley–Hamilton theorem (Wallin & Johansson 2002), or Equation 16 can be solved as a set of three simultaneous equations for $\omega^S_i$. In two dimensions, $w^S$ is in the $x_3$ direction, so that $\Omega^S = W^{SS}$.

This surrogate for curvature and rotation, then, is introduced into models. It represents a unification of rotation and curvature because, in a rotating coordinate system, $\Omega^S$ includes the system rotation tensor. A model that was devised for system rotation is made into a model for both rotation and streamline curvature by replacing coordinate frame rotation $\Omega^F$ by rotation of the rate of strain $\Omega^S$; that is, just substitute $\Omega^F \to \Omega^S$.

For instance, the absolute rotation tensor is

$$\Omega^A = \Omega + \Omega^F,$$

where $\Omega$ is the vorticity tensor relative to a rotating frame. $\Omega^A$ appears in some models. One simply redefines $\Omega^A$ as

$$\Omega^A = \Omega + \Omega^S.$$ 

In Reynolds stress models, frame rotation enters via

$$\Omega^* = \Omega + C\Omega^F,$$

where $C_\epsilon$ is a model constant (Speziale et al. 1991, Wallin & Johansson 2002). The analogy between curvature and rotation is invoked by replacing $\Omega^F$ with $\Omega^S$ to obtain

$$\Omega^* = \Omega + C\Omega^S.$$ 

Hellsten (1998) sensitized the $k - \omega$ model to coordinate system rotation via a function of the Bradshaw number [he simply replaced a constant value of $C_{\omega_0} = 3/40$ by $3/40(1 + 3.6Br)$]. An equilibrium solution in rotating homogeneous shear flow shows this to stabilize turbulence for rotation numbers above 0.24 or below $-1.24$ (Durbin 2011), which are approximately in agreement with experiments. Khodak & Hirsch (1996) proposed the invariant definition

$$Br = \frac{|\Omega^A|(|\Omega^A| - |S|)}{|S|^2}$$

of the Bradshaw number. The replacement, stated in Equation 17, makes this a curvature Bradshaw number. Arolla & Durbin (2013) applied the method of Equation 17 to other models that incorporate system rotation via invariants of the velocity gradient. It was validated by computations of rotating, curved, and swirling flows.
Shur et al. (2000) devised a function of the parameter

\[ \tilde{\rho} = \frac{|S|^2 \Omega \cdot W_{SS}}{(|S|^2 + |\Omega|^2)^2} \]

to enhance or suppress production of eddy viscosity, depending on whether curvature was stabilizing or destabilizing. It was validated in rotating and curved channels.

2.3.2. Buoyancy. Methods for incorporating buoyancy in stratified shear flow are well established. Stability functions are introduced to parameterize mixing as a function of the Richardson number. The stability functions can be derived from second moment closure. Early analyses used overly simple pressure–strain models (Mellor & Yamada 1982). Canuto et al. (2001) developed an updated version. Equilibrium analysis of stratified shear flow shows how the stability functions, or simply the equilibrium curve of $Sk/\varepsilon$, suppress turbulent energy above a critical Richardson number (Canuto et al. 2001, Ji & Durbin 2004). The critical Richardson number is not entirely certain—values range from 0.25 to 1.0.

The effects of buoyancy have been especially hard to capture in the problem of flow along a vertical, heated wall. Let $\theta$ stand for temperature. The usual scalar diffusion hypothesis,

\[ -u \theta = \alpha T \nabla \Theta \]

gives quite poor results. More generally, one can propose a tensorial relation

\[ -u \theta = \alpha_T \cdot \nabla \Theta \]

The generalized gradient diffusion hypothesis asserts that $\alpha_T$ is proportional to $uu$. However, this is incorrect: $uu$ is a symmetric matrix and the diffusion matrix is decidedly not symmetric. That is known from experiment and direct numerical simulation (DNS), and also can be deduced from the transport equation for $u \theta$. Kenjereš et al. (2005) proposed an algebraic closure of that transport equation in the form

\[ \left( I - c_1 b + c_2 T \nabla U \right) \cdot \vec{u} = -c_3 \left( uu \cdot \nabla \Theta + c_4 \beta g \theta^2 \right), \]

where $b$ is the Reynolds stress anisotropy tensor and $T$ is the turbulence timescale. This differs from previous models by adding the anisotropic relaxation term $c_1 b$. If the matrix on the left is inverted, the first term on the right becomes a diffusion tensor. The velocity gradient on the left makes the diffusion tensor symmetric. The second term on the right provides a buoyant contribution to heat flux. Kenjereš et al. (2005) showed how this formulation captures buoyancy more accurately. Suga & Abe (2000) simply posed an asymmetric $\alpha_T$, with asymmetry introduced by the vorticity tensor. They also included an anisotropy term and showed that it improves agreement to data. A more extensive discussion of scalar flux closure options can be found in Hanjalić & Launder (2011).

3. ADDING TRANSITION

The computational domain often includes a laminar to turbulent transition. That may have either a primary or a secondary effect on the turbulent flow prediction. Whether or not a transition occurs may determine whether the boundary layer is separated or attached. Or, the length of the surface in which the boundary layer is laminar might be needed for overall drag prediction.

The laminar to turbulent transition is classified into bypass and orderly modes (Zaki 2013). Bypass dominates when an attached boundary layer is subjected to free-stream disturbances, such as ambient turbulence or incident wakes. It is mediated by streaks in the laminar boundary layer. Orderly transition is mediated by instability waves. It is seen under very quiet streams, in strong adverse pressure gradients, in boundary layers with cross-flow, or in supersonic flows. Instability waves also dominate in separated flow and transition often occurs rapidly after separation. In some
BOUNDARY LAYER TRANSITION

The laminar to turbulent transition in boundary layers can be envisioned as a rise in the average level of skin friction, occurring at some downstream distance. Then the basic objective of a model is to predict where that rise occurs. The levels before and after the rise are given by the laminar and turbulent values.

Intermittency: the fraction $0 \leq \gamma \leq 1$ of the time that the flow is turbulent.

cases, both streaks and instability waves arise and interact to cause mixed-mode transition (Liu et al. 2008).

Transition in boundary layers has traditionally been accommodated by data correlations. The correlation specifies a transition point, usually stated as the momentum thickness Reynolds number at transition (see the sidebar titled Boundary Layer Transition). The solution is laminar up to transition; the turbulence model is invoked post-transition. Modern, applied CFD requires a more flexible approach.

Adding the transition to Reynolds-averaged closures has presented a great challenge to modelers. Three approaches have been devised: (a) rely on the closure model to transition from a laminar to a turbulent solution, (b) use a data correlation to decide when to switch from laminar to turbulent flow, or (c) devise additional transport equations to represent the transition. In this last category, two approaches have been pursued. The first invokes an equation for an intermittency function $\gamma(x,t)$, which is 0 in laminar flow and 1 in turbulent flow. The second is to develop a separate equation for the kinetic energy of velocity fluctuations $k_L$ in the laminar region upstream of transition and to model how it is transferred into turbulent kinetic energy.

The oldest approach is to switch from a laminar to a turbulent computation at a prescribed transition point. Abu-Ghannam & Shaw (1980) proposed a correlation for bypass transition that becomes

$$R_{\theta_{tr}} = 163 + e^{6.91 - T_u}$$

under zero pressure gradient. $T_u$ is the turbulence intensity as a percentage, $100\sqrt{\overline{u'^2}} / U$, measured in the free stream. The transition is specified to occur where the local momentum thickness Reynolds number exceeds this critical value. The full correlation accounts for pressure gradients as well. Suzen & Huang (1999), among others, have proposed variants of the Abu-Ghannam & Shaw correlation. However, it can be difficult, if not impossible, to compute momentum thickness in a complex flow.

3.1. Reliance on the Turbulence Model

It is not uncommon to encounter regions of purely laminar flow or buffeted laminar flow in applications. For instance, turbine blades often operate at low enough Reynolds numbers to encounter significant portions of laminar flow on their surface, but the blades are subjected to external turbulence, so their boundary layers are better described as buffeted laminar layers (Hodson & Howell 2005). In such instances, there are irregular fluctuations in the laminar region, and the rest of the flow is turbulent. The overall flow calculation must be done with a turbulence model. Perhaps one can dispense with a transition model and rely on the turbulence closure to return a nearly laminar solution in the buffeted region and to predict transition. Is this a plausible approach?

On the face of it, the answer is no. Closures are developed for fully turbulent conditions and calibrated with turbulence data. However, most transport equation models do converge to a laminar...
solution at low Reynolds number and to a turbulent solution at sufficiently high Reynolds number; in this sense, the model equations evidence a transition between laminar and turbulent solution branches. So relying on the closure model may not be entirely irrelevant, but neither is it entirely justified. Savill (1999) and Westin & Henkes (1997) explored whether low–Reynolds number \( k-\varepsilon \) models can imitate bypass transitions. In some cases they found satisfactory transitional behavior. Pecnik et al. (2005) found satisfactory boundary layer transitions in calculating a turbine stage using the \( v^2-f \) model, as did Wu & Durbin (2000) for wake-induced transitions. Wilcox (1994) suggested that low–Reynolds number modifications could enable \( k-\omega \) to be used in transitional flow. (It should be remarked that the term low Reynolds number refers to the near-wall region of fully turbulent flow.)

However, the switch from laminar to turbulent solutions in the standard forms of the \( k-\omega \) model, \( k-\omega \) shear stress transport (SST) model, and S-A model occurs at a far too low Reynolds number. Despite some successes, the turbulence model is not, in fact, a method to rely on. When accurate predictions of the laminar and transitional regions are needed, the turbulence model must be supplemented explicitly by a method to predict transition.

### 3.2. Intermittency Equation

After the transition criterion \( R_{\alpha} > R_{\alpha T} \) (e.g., Equation 20) is met, Dhawan & Narasimha (1958) proposed that turbulent energy could be ramped up by an intermittency function, which increases across the transition region from the laminar value of \( 0 \) to the turbulent value of \( 1 \). They suggested a half-Gaussian shape: \( \gamma = 1 - \exp[-(x - x_T)^2/\ell_{tr}^2] \), where \( x > x_T \). Lodefier et al. (2004) derived the form of an intermittency transport equation from that Gaussian shape:

\[
D_\gamma \frac{d\gamma}{dt} = \frac{2(2\gamma - 1)\sqrt{\nu}}{\ell_{tr}} + \nabla \cdot [(\nu + \nu_T)\nabla \gamma],
\]

where \( \ell_{tr} \) is a transition length. This is a base for more elaborate formulations, which include a sink term as well a source:

\[
D_\gamma \frac{d\gamma}{dt} = P_\gamma - E_\gamma + \nabla \cdot [(\nu + \nu_T)\nabla \gamma].
\]

Langtry & Menter (2005) and Menter et al. (2006) included a second equation in their \( \gamma - R_{\alpha T} \) model. The \( R_{\alpha T} \) equation diffuses a data correlation from the free stream into the boundary layer. Then the model predicts both transition onset and intermittency in the transition region. Langtry & Menter (2005) rewrote the source minus the sink term on the right side of Equation 22 as

\[
2\left|\frac{S}{N}\right|(\gamma - 1)\sqrt{\gamma}f_{\text{onset}}f_{\text{length}} + (0.06 - 50.0\gamma)F_{\text{turb}}/\Theta_1 \gamma,
\]

where \( f_{\text{onset}} \) and \( F_{\text{turb}} \) are complicated, highly empirical functions. The model was published without the formula for \( f_{\text{length}} \), which was proprietary. Suluksna et al. (2009) developed the missing information and published a full statement of this model.

An additional objective was to make the transition depend on local variables, rather than the integral boundary layer thickness \( \Theta_1 \), as in Equation 20. To that end, the local parameter

\[
R_{\alpha} = \frac{|S|d^2}{2.193\nu},
\]

where \( d \) is distance from the wall, is used in place of \( R_{\alpha} \).

The intermittency method is predicated on the underlying RANS model predicting early transition. Then transition is suppressed by multiplying either production or the eddy viscosity by \( \gamma(x) \leq 1 \). As \( \gamma \) rises from zero to one, the model is switched on.
A question arose as to whether transition can be predicted by a single intermittency equation. Durbin (2012) and Ge et al. (2014) developed an intermittency equation that predicts the transition with neither an adjoined $R_T$ equation nor a data correlation per se. In this approach, transition is controlled by both a source/sink and a diffusion of free-stream turbulence into the boundary layer. The source/sink is written

$$F_{\gamma}|\Omega| (\gamma_{\text{max}} - \gamma) \sqrt{\gamma} - G_{\gamma} F_{\text{turb}}|\Omega| \gamma^{1.5},$$

where $F_{\gamma}$, $G_{\gamma}$, and $F_{\text{turb}}$ are complicated functions of the nondimensional, Galilean-invariant parameters

$$R_t \equiv \nu_T |\Omega|, \quad T_\omega \equiv R_t |\Omega| / \omega, \quad R_\nu \equiv d^2 |\Omega| / \sum 188 v,$$

where $d$ is wall distance. $\sigma_{\gamma}$ was set to 0.2 in Equation 22 to control diffusion into the boundary layer.

Menter et al. (2015) subsequently proposed a single $\gamma$ equation, which they characterized as a reduction of the $\gamma - R_T$ model. However, it invoked Galilean-invariant parameters similar to those above, rather than the Galilean-dependent parameters of their earlier model.

Although the $\gamma - R_T$ model is devised to be used with the two-equation, $k - \omega$ turbulence model, Medida & Baeder (2011) devised a reduced version for use with the one-equation, S-A eddy viscosity transport model.

The model of Ge et al. (2014) was extended to roughness-induced transitions by Ge & Durbin (2015). They used a displaced origin as a function of wall roughness to modify the model. In order to represent roughness, Dassler et al. (2010) added a transport equation for roughness amplification to the model of Langtry & Menter (2005).

Wang & Fu (2011) proposed a single $\gamma$ equation of the form of Equation 22, without a separate data correlation equation, for supersonic and hypersonic transition. They introduced timescales to represent orderly transitions by breakdown of Mack waves.

A disconcerting aspect of this type of model is that the sink term $E_{\gamma}$ must establish a laminar boundary layer. The intermittency equation is solved everywhere, including the laminar boundary layer; a region of $\gamma = 0$ must be created by its sink term in order to define the laminar zone. That is unlike a simple data correlation, for which one computes a laminar boundary layer and then switches on the turbulence model.

### 3.3. Laminar Fluctuations

A second approach has a closer connection to the phenomenon of transition. Mayle (1991) proposed that the transition could be modeled through consideration of laminar fluctuation energy. That might at first seem a perverse notion: a laminar turbulent kinetic energy, denoted $k_L$. On second consideration, it is not so odd: The laminar fluctuating energy can be construed to represent instability waves or the streaks seen prior to the bypass transition (Zaki 2013). Such ideas are attractive, even though streaks and instability waves do not enter directly into these models. Moreover, streaks and instability waves are only precursors to the transition. In isolation, they are not the cause of turbulence; it is spawned when they break down through secondary processes.

The key elements of the model are the production of laminar fluctuations and the transfer of energy from laminar to turbulent fluctuations. Walters & Leylek (2004) and Walters & Cokljat (2008) proposed the form

$$D_t k_L = 2 v_T |S|^2 - R - \varepsilon_L + \nabla \cdot (\nu \nabla k_L),$$

where $D_t$ is the time derivative. This equation captures the production of laminar fluctuations due to the nonlinear terms in the momentum equations and the dissipation of laminar energy.

$$\varepsilon_L = \nu \nabla k_L,$$

which represents the dissipation of laminar energy into turbulent energy. The term $R$ represents the production of turbulent energy due to the nonlinear terms in the momentum equations.

$$k_L = k_L - 2 v_T |S|^2,$$

which is the transport equation for laminar fluctuations. This equation is solved in the laminar boundary layer, where the sink term $E_{\gamma}$ must establish the laminar zone. A region of $\gamma = 0$ must be created by its sink term in order to define the laminar zone. That is unlike a simple data correlation, for which one computes a laminar boundary layer and then switches on the turbulence model.

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Hybrid: a combination of eddy-resolving simulation and a RANS closure.

The term \( R \) in Equation 23 represents breakdown of laminar fluctuations into turbulence. The same term, with a positive sign, is added to the turbulent energy equation:

\[
D_t k = 2\nu_T|S|^2 + R - \varepsilon + \nabla \cdot \left( \nu + \frac{\nu_T}{\sigma_k} \nabla k \right).
\]

Walters & Leylek (2004) invoked \( k - \varepsilon \) for the turbulence model and Walters & Cokljat (2008) invoked \( k - \omega \).

Lopez & Walters (2016) proposed a variant of the laminar kinetic energy approach in which a velocity variance \( \overline{\omega^2} \) is used instead of \( k_L \). This variable was motivated by the role of wall-normal fluctuations in causing transitions. However, in fully turbulent flow it becomes the turbulent kinetic energy, \( \overline{\omega^2} = k \). The new model is similar to that of Walters & Cokljat (2008), but fixes seriously inaccurate predictions of free-shear flows by the earlier model.

Kubacki & Dick (2016) reduced the laminar fluctuation model to a simple algebraic formula. They were able to predict transition on turbine blades.

Lardeau et al. (2004) combined a laminar fluctuation model with a nonlinear constitutive model (Section 2.2) based on \( k - \varepsilon \). They represented the total fluctuating energy as \( \gamma k_T + (1 - \gamma) k_L \). However, their model retains a data correlation for transition and the semi-Gaussian form of Dhawan & Narasimha (1958) for the intermittency function.

Transition prediction is safest when it is rooted in a data correlation. At the present time, the \( \gamma - R_T \) model is the most successful for that reason. Laminar fluctuation models have not achieved the same level of predictive consistency.

4. USING RANS AS A HYBRID

Often RANS does not predict mixing in detached shear layers well. Eddy-resolving simulation is more accurate but more costly. Is there a compromise between the expense of eddy simulation and the efficiency of RANS modeling? Hybrid methods are motivated by this question. They enable eddy-resolving simulation on grids that are far too coarse for DNS and generally too coarse for wall-resolved LES. The grids at least have to be adequate for Reynolds-averaged computation. Fröhlich & Von Terzi (2008) extensively reviewed hybrid simulation.

When it is used as a hybrid, the RANS model becomes dominant in some parts of the flow—especially near walls—while an eddy-resolving simulation, with a subgrid model, prevails elsewhere. The terms zonal and nonzonal (Walters et al. 2013) are sometimes used. Zonal means that a region of the computational domain is prescribed in advance as the RANS zone, and the remainder is prescribed as the eddy simulation zone. Most work of this persuasion has come from the LES community, which has approached it as appending a RANS zone onto an LES (see Bose & Park 2018 in this volume).

Here we are concerned only with nonzonal methods. Most work of that persuasion has come from the RANS community, modifying the closure model to allow turbulent dynamics to occur. There are no RANS and eddy simulation zones per se.

One also can refer to seamless methods (Hanjačić & Kenjereš 2008) and automatic, seamless methods (Slotnick et al. 2014). It is unclear whether the terms nonzonal and seamless are distinct in prior usage: We will distinguish seamless methods as those which use only a single closure.
model through the domain. The alternative is a combination of models that acts like RANS or LES, depending on which is dominant.

In a nonzonal method, one computes 3D, time-evolving solutions to the Navier–Stokes equation

\[
\frac{DU}{Dt} = -\frac{1}{\rho} \nabla p + \nu \nabla^2 U - \nabla \cdot \tau \tag{25}
\]

in the entire domain, with a model for the unclosed stress \( \tau \). Various substitutions for the unclosed stress cause the computation to emulate Reynolds-averaged or turbulence-resolving solutions. In the latter case, various substitutions cause the computation to be either an LES or a hybrid simulation.

The hybrid simulation adds dynamic equations for the closure. For example, for \( k - \omega \), they are

\[
\frac{Dk}{Dt} = 2v_T |S|^2 - C_{\mu} k \omega + \nabla \cdot [(\nu + \sigma_k(k/\omega))\nabla k], \tag{26}
\]

\[
\frac{D\omega}{Dt} = 2C_{\omega} |S|^2 - C_{\omega2} \omega^2 + \nabla \cdot [(\nu + \sigma_\omega(k/\omega))\nabla \omega].
\]

The closure of Equation 25 is

\[
-\tau = 2v_T S - \frac{2}{3} k I, \tag{27}
\]

where \( v_T = k/\omega \) and \( S \) is the rate of strain tensor, \( S_{ij} = \frac{1}{2}(\partial_j U_i + \partial_i U_j) \). More correctly, this is the RANS formulation; in hybrid simulations some aspect of Equation 26 or Equation 27, or both, is modified to enable eddying.

### 4.1. Blended Formulations

The nonzonal, but not seamless, approach is to stitch together a RANS and an LES subgrid scale (SGS) model (Fröhlich & Von Terzi 2008). A function of the ratio of grid dimension to turbulence scale \( f(\Delta/\Lambda) \) is introduced such that \( f(\infty) = 0 \) and \( f(0) = 1 \). The stress models are combined as

\[
\tau = (1 - f)\tau^{\text{RANS}} + f\tau^{\text{SGS}}. \tag{28}
\]

The flow simulation methodology model (Speziale 1998, Fasel et al. 2002) is of this form, with \( \tau^{\text{SGS}} = 0 \). It is described as a blend of RANS and DNS. However, hybrid simulation is meant for grids too coarse for DNS, so it could just as well be described as a blend of RANS with implicit LES—which means that numerical diffusion acts as \( \tau^{\text{SGS}} \). An algebraic stress model (see Section 2.2) is adopted by Fasel et al. (2002) for \( \tau^{\text{RANS}} \) along with the \( k - \varepsilon \) model. The Kolmogoroff scale \( \eta = (\nu^3/\varepsilon)^{1/4} \) is used as the turbulence length scale and the blending function is

\[
f = \exp(-0.004\Delta/\eta).
\]

Walters et al. (2013) and Bhushan & Walters (2012) evaluated the interpolation function dynamically by taking the inner product of the rate of strain tensor \( S_j \) with Equation 28 and solving

\[
1 - f = \frac{\langle \tau - \tau^{\text{SGS}} \rangle : \langle S \rangle}{\langle \tau^{\text{RANS}} - \tau^{\text{SGS}} \rangle : \langle S \rangle}. \tag{29}
\]

The stress difference in the numerator is the resolved stress, as calculated during the simulation. This formula is clipped so that \( f \) is constrained to lie between 0 and 1. Walters et al. (2013) invoked implicit LES; that is, they relied on numerical diffusion in place of an explicit model for \( \tau^{\text{SGS}} \). The eddy viscosity constitutive formula in Equation 27, with the \( k - \omega \)-SST eddy viscosity, was used for the RANS stress. Bhushan & Walters (2012) invoked the dynamic Smagorinsky, subgrid model for \( \tau^{\text{SGS}} \) and the S-A RANS model.
Uribe et al. (2010) attributed the idea for their two-velocities model to Schumann (1975). It is of the form of Equation 28 with

\[
\tau_{ij}^{\text{RANS}} - \frac{1}{2} \delta_{ij} \tau_{kk}^{\text{RANS}} = 2 \nu_T \langle S_i \rangle,
\]

\[
\tau_{ij}^{\text{SGS}} - \frac{1}{2} \delta_{ij} \tau_{kk}^{\text{SGS}} = 2 \nu_{\text{SGS}} (S_j - \langle S_j \rangle),
\]

where the angle brackets represent an ensemble average: It is computed as a running average over about 10 eddy turnover times. For \( \nu_T \) they use the \( \phi - f \) model (Laurence et al. 2004), in which the eddy viscosity is \( \nu_T = C_\phi \phi k T \), where \( T \) is the turbulence timescale. The subgrid viscosity \( \nu_{\text{SGS}} \) is the Smagorinsky form, \( (C_s \Delta)^2 |S - \langle S \rangle| \). In this case, \( \Delta \) is the integral scale, \( \phi k^{3/2} / \epsilon \). In this formulation, at any point in the flow domain, two different velocity fields, \( U' \) and \( \langle U' \rangle \), coexist. This allows automatic, dynamic switching between RANS and LES as a function of the grid resolution compared to \( \Lambda \). Xiao & Jenny (2012) took the two-velocity concept a step farther, and computed simultaneous, coupled RANS and LES fields on two different grids.

All these models are validated in channel flow, boundary layers, or more complex geometries, and are shown to be effective. In general, hybrid models improve the prediction of reattachment lengths.

### 4.2. Seamless Methods

Another class of nonzonal method—which satisfy the present criterion for being seamless—modifies coefficients of a RANS model via a function of the ratio of grid dimension to turbulence scale \( \Delta / \Lambda \). These include the partially integrated transport model (PITM) (Chaouat & Schiestel 2005, 2012) and the partially averaged Navier–Stokes (PANS) model (Girimaji & Abdol-Hamid 2005). In these methods, \( \Lambda \) is the integral scale \( k^{3/2} / \epsilon \). They are framed around the destruction term in the \( \epsilon \) equation. This has a coefficient \( C_{\epsilon 2} \) equal to 1.92. In PITM, it is replaced with

\[
C_{\epsilon 2}^* = C_{\epsilon 2} + f(\Delta / \Lambda) (C_{\epsilon 1} - C_{\epsilon 2}),
\]

where the constant \( C_{\epsilon 1} = 1.44 \) is the coefficient of the production term in the \( \epsilon \) equation. Thus, when the grid is coarse \( C_{\epsilon 2}^* \) becomes \( C_{\epsilon 2} \); on a fine grid, it becomes \( C_{\epsilon 1} \).

The manner in which this allows eddying is not entirely clear. Chaouat & Schiestel (2005) argued that when the RANS formulation is acting as a subgrid model, it is transferring resolved kinetic energy to subgrid energy and dissipating that energy. When \( C_{\epsilon 2} = C_{\epsilon 1} \), an equilibrium solution exists in which production equals dissipation. Then the balance between production and dissipation makes the subgrid model directly dissipative, as needed for LES.

Another, more direct explanation, is that reducing \( C_{\epsilon 2}^* \) increases dissipation \( \epsilon \), which in turn reduces \( k \). In combination, these reduce the eddy viscosity \( C_{\epsilon 2}^* k^2 / \epsilon \), enabling the LES mode. PANS operates similarly to PITM, although it modifies both the production and destruction terms in the \( \epsilon \) equation to enable a balance where production equals dissipation, and to reduce the eddy viscosity.

Fadai-Ghotbi et al. (2010) invoked the PITM method in conjunction with an elliptic blending Reynolds stress model (see Section 2.1). In PITM and PANS, \( f(\Delta / \Lambda) \) originates as a ratio \( r \) of modeled to total turbulent kinetic energy, which is developed into a function of \( \Delta / \Lambda \). Fadai-Ghotbi et al. (2010) reverted to the initial idea, basing \( r \) on a ratio of the modeled to the actual, computed total kinetic energy. This was done because the PITM formulation for \( f(\Delta / \Lambda) \) inhibited eddy formation in channel flow.

These methods have been validated by computer simulation and compared successfully to experimental or DNS data. In some cases the validation is in a very simple geometry, like channel
flow or a flat plate boundary layer, and it is unclear how robust the method is. None of these methods has the widespread use enjoyed by the next class of methods.

4.3. Detached-Eddy Simulation

Hybrid RANS/LES models are considered to have promise for industrial CFD applications. The idea is to rely heavily on a closure model in the near-wall part of attached boundary layers and on eddy-resolving simulation in regions away from the surface. The most popular seamless hybrid is detached-eddy simulation (DES) (Spalart 2009). Although DES has some similarity with the methods reviewed above, more than them, it is seen in the vein of enabling a coarser near-wall mesh than would be required for LES.

It appears incongruous for a hybrid to be both an ensemble average (RANS) and a realization (LES). DES is better described as a length scale formulation that adopts either a RANS formula, or a grid dimension, as in LES. It is, everywhere, a realization that has to be averaged to obtain statistics. Significant levels of fluctuations usually are seen, even in parts of the domain that are labeled RANS regions. In delayed DES (DDES), a shielding function \( f_d \) is introduced, and \( f_d \sim 0 \) is called the RANS region, as shown in Figure 2.

Figure 3 portrays instantaneous and averaged \( Q \) contours on a 10 million–cell RANS mesh. This is a simulation of air flow in a prototype combustor atomizer, computed by adaptive DDES (Ryon 2016). It illustrates that DES resolves turbulent eddies. The averaged field is computed by ensemble averaging the simulation. Figure 3b shows a purely RANS computation and includes a greater length of the domain than Figure 3a,b. The ensemble average is similar to the RANS computation, but at considerably greater computational cost. The cost is warranted if accuracy is improved, or if space-time turbulence structure is needed. Both justifications apply to this atomizer study.

DES originated as the simple formula \( \ell = \min[\ell_{\text{RANS}}, C_{\text{DES}}\Delta] \), where \( \Delta \) is the grid spacing (Spalart 2000). So the RANS region is that in which the first term is smallest. In most cases, the
RANS region is next to walls; however, DES is not a wall model per se. Generally, where the grid is coarse compared to the requirements for LES, DES will adopt the RANS length scale; where the grid is fine, it will adopt grid spacing.

Since it was first proposed, DES has undergone considerable revision, as reviewed by Spalart (2009). An early fault was the occurrence of grid-induced separation, which was attributed to a shortage of resolved stress in intermediate regions where eddy viscosity is decreasing, but resolved stresses have not developed fully. In order to alleviate this, Spalart et al. (2006) introduced the shielding function $f_d$. Also, the notion that the method was only for simulating eddying in detached shear layers gave way to a more general view of hybrid, eddy-resolving simulation—so that the D (detached) in DES is not to be taken literally. This more general view initially met a hurdle. Attempts at channel flow simulation (Piomelli et al. 2003) resulted in two mismatched log layers: one from the RANS branch and the other from the LES branch. This anomaly was termed log-layer mismatch and led to other modifications of DDES, such as improved DDES (see Spalart’s 2009 review) and vorticity-aligned length scales (Shur et al. 2015).

Originally, DDES was developed for the S-A, eddy viscosity transport model. In that case, the RANS length scale is the wall distance, which enters a dissipation term. The method was applied by Strelets (2001) to the two-equation, $k – \omega$ model. He followed precedent, rewriting the dissipation term $C_{\mu}k/\omega$ as $C_{\mu}k^{1/2}/\ell$ in the $k$-component of Equation 26 in order to phrase it in terms of a dissipation length scale, and then used the DDES length scale formulation of Equation 32. Hanjalić & Kenjereš (2008) described a similar approach due to Hadžibedinić (2005), based on the dissipation term in the $k – \varepsilon$ model. Gritskевич et al. (2012) adopted the Strelets formulation directly. Ashton et al. (2012) developed a variant that used the elliptic relaxation $\phi – f$ model (see Section 2.1). Jee & Shariff (2014) developed a version based on the elliptic relaxation $v^2 – f$ model. In addition to the dissipation length scale, they introduced the timescale $\ell_{DES}/\sqrt{K}$, and obtained especially good results for flow over an airfoil.

Reddy et al. (2014) developed a different approach, writing the eddy viscosity as $\nu_T = \ell^2\omega$. Correspondingly, the production term in the $k$-component of Equation 26 becomes $2\nu_T|S|^2 = 2\ell^2\omega|S|^2$. Hence, in this formulation, the lower limit on dissipation in the Strelets (2001) approach
is replaced by an upper limit on production. Again, the DDES length scale (Equation 32) was invoked. Importantly, the switch to the $\ell^2 - \omega$ formulation facilitated the development of adaptive DDES, as described below.

The DDES length-scale model is

\[
\ell_{\text{DDES}} = h_{\text{RANS}} - f_d(r_d) \max(0, h_{\text{RANS}} - \ell_{\text{ES}}),
\]

\[
r_d = \frac{k/\omega + \nu}{\kappa^2 d_w^2 \sqrt{|S|^2 + |\Omega|^2}},
\]

\[
h_{\text{RANS}} = \sqrt{k/\omega}, \quad \ell_{\text{ES}} = C_{\text{DES}} \Delta_1,
\]

where Strelets (2001) used $\Delta = h_{\text{max}}$ and Reddy et al. (2014) use $\Delta = V^{1/3}$. $V$ is the cell volume, $h_{\text{max}}$ is the maximum cell dimension, and $d_w$ is the wall distance. The shielding function $f_d$ ensures a near-wall RANS region on any grid. It has the limits $f_d(\infty) = 0$ and $f_d(0) = 1$.

Yin et al. (2015) and Yin & Durbin (2016) made the $\ell^2 - \omega$ model adaptive by using the dynamic procedure of LES (Lilly 1992),

\[
C_{\text{dyn}}^2 = \max\{L_{ij} M_{ij} / 2 M_{ij} M_{ji}, 0\},
\]

in which the test filter stress and rate of strain,

\[
L_{ij} = -\hat{u}_i \hat{u}_j + \hat{\alpha} \hat{S}_{ij},
\]

\[
M_{ij} = \Delta^2 \hat{\alpha} \hat{S}_{ij} - \Delta^2 \hat{\alpha} \hat{S}_{ji},
\]

are computed from the resolved velocity field $\mathbf{u}$ by local averaging—indicated by the caret. The DES coefficient is given by

\[
C_{\text{DES}} = \max(C_{\text{lim}}, C_{\text{dyn}}).
\]

The lower bound $C_{\text{lim}}(b_{\text{max}}/\eta)$ prevents spuriously low values of $C_{\text{DES}}$ on grids that are too coarse for the dynamic procedure. It compares grid size to an estimated Kolmogoroff scale, $\eta = (v^3/\varepsilon_{\text{est}})^{1/4}$, where $\varepsilon_{\text{est}} = k\omega$. As $b_{\text{max}}/\eta \to \infty$, $C_{\text{lim}} \to 0.12$ is the default, coarse grid limit. When $b_{\text{max}}/\eta \ll 20$, $C_{\text{lim}} \sim 0$, so $C_{\text{DES}}$ is constrained only to be nonnegative.

An indirect effect of the dynamic, local evaluation of $C_{\text{DES}}$ is that the shielded region shrinks when the grid is relatively fine (Figure 4a). $f_d \sim 1$ is the turbulence-resolving region. The $f_d$ curve starts to rise when $y_+ < 10$; the turbulence already is resolved in the log layer where $y_+ > 40$. The rational limit, that DES $\to$ wall-resolved LES as the grid is refined, can be accomplished by adaptive DES (Yin & Durbin 2016).

The adaptive model reverts to nonadaptive behavior on coarse grids (Figure 4c,d). The non-zonal character of DES is also illustrated by the $f_d$ curves: The RANS region is not of a fixed thickness, varying from one case to the next.

5. DATA-DRIVEN MODELING

Large sets of Reynolds-averaged data have been created by DNS and advanced measurement methods. It has long been speculated that these might lead to improvements of turbulence models. However, the large data sets have mostly been used in the same manner as older, more limited data sets: that is, they are plotted along with model predictions to assess model accuracy by comparison. Certainly, channel flow DNS data have been especially useful for compute-and-compare assessment and development of models. However, there is a persistent sense that data-driven methods have had too little impact on closure modeling.

Parneix et al. (1998) proposed that DNS data might be used more comprehensively to assess models. Some terms in the governing equations can be taken from data, and the model’s ability to
predict others can be assessed. For instance, $\varepsilon(x)$ can be taken from a DNS field of data, removing the suspicious $\varepsilon$ equation from the model. Actually, Parneix et al. (1998) concluded that the $\varepsilon$ equation was not the cause of poor predictions in flow over a backward-facing step. Raiesi et al. (2011) applied a similar method to LES data sets. As one example, they used a field of $k$ from LES data and solved the $\omega$ equation of the $k-\omega$ model. Not surprisingly, using accurate $k$ deteriorates the accuracy of the $k-\omega$ model (one could say that the $k$ of the $k-\omega$ model should not be interpreted as the turbulent kinetic energy). Raiesi et al. (2011) also took $k$ and $\varepsilon$ from LES and solved the $\zeta$ and $\phi$ equations of the $\zeta-\phi$ model. Ling et al. (2013) used experimentally measured velocity and thermal fields and obtained an eddy diffusivity by optimizing the fit between model and data. Regrettably, these studies have not led to insights into improving closure models.

Recently, an approach termed inverse modeling was advocated by Dow & Wang (2010, 2011), Duraisamy et al. (2015), and Parish & Duraisamy (2016). It has been applied to transition (Duraisamy & Durbin 2014) and turbulence models. Dow & Wang (2010) provided a simple exposition of the method for plane channel flow.

Inverse modeling evolved out of adjoint-based, optimization methods for aerodynamic design. A fundamental juggernaut to using data more comprehensively has been the ambiguous connection between model variables—$\omega(x)$, $\nu_1(x)$, $\gamma(x)$, etc.—and physically measurable quantities. Model variables are an invention of the closure scheme. Inverse modeling treats them as parameters to be found by minimizing a cost function. For instance, the cost function may be the weighted mean-squared error

$$E = \sum_i |Q^i - Q_{data}^i|^2 w_i$$

of a predicted variable $Q$, relative to an array of data $Q_{data}$. If a model variable $\phi(x)$ is to be found, it is treated as a parameter field and

$$\min_{\phi(x)} E$$

is solved, where $n$ includes all grid points. In the approach proposed by Duraisamy et al. (2015), instead of directly finding the turbulence variable, a corrective coefficient $\alpha$ is inserted into its governing equation. One writes

$$\frac{D\phi}{Dt} = \alpha(x) \times \text{Source} - \text{Sink} + \text{Diffusion}$$
and \( \min_{\alpha(x)} E \) is solved, starting from the initial guess \( \alpha \equiv 1 \). Thereby, one obtains a field of the variable of interest, e.g., \( \omega(x) \). Hence, data on \( \omega(x) \) are extracted from data on the flow field by solving the \( \omega \) equation with the optimal correction \( \alpha \). The method is illustrated by Figure 5, in which an optimized fit to skin friction data is used to extract an eddy viscosity profile.

Inverse modeling is attractive because data for model variables are extracted by using those variables in their intended context. In a RANS computation, predictions of quantities like skin friction or the velocity field are global consequences of model variables. Therefore, as the cost function is minimized, at each iteration, the full RANS equations must be solved.

In the adjoint method, the gradient of the cost function is found by solving linear, adjoint equations (Giles & Pierce 2000). The adjoint solution functions as sensitivity coefficients that weight regions where the corrective function has its largest impact on the discrepancy between prediction and input data.

To date, this has been a method to extract data on model variables. It has yet to lead to improved prediction methods. Extracted data are functions of position in a given geometry \( \alpha(x) \). Research is in progress to convert this into model improvements by correlating \( \alpha \) with coordinate independent parameters, using methods of machine learning such as neural nets or kernel regression (Tracey et al. 2013, Parish & Duraisamy 2016).

### 6. OUTLOOK

Over the years, the notion that closure modeling will be replaced by eddy-resolved simulation has been advanced and abandoned (Laurence 2002, Hanjalić 2005, Hanjalić & Kenjereš 2008, Spalart 2010, Slotnick et al. 2014). Mark Twain is often reported to have quipped, “reports of my death have been greatly exaggerated.” It is a matter of extrapolation, not prophecy, that RANS closure models will remain an essential element of applied CFD into the foreseeable future. Their marriage with eddy-resolving simulation, under the heading of hybrid methods, adds to their vitality. Hybrid simulation is expanding into new niches where it provides either improved accuracy or space and time fluctuations that are not available from pure RANS solutions for ensemble-averaged fields. At the same time, niches for pure RANS analysis are also expanding as industry and researchers find new applications. Advances in modeling have not kept pace with this burgeoning use.
The basic limitations—that moment equations are unclosed and that closure is accomplished by empiricism—are inescapable: Averaging destroys information and prohibits mathematical rigor. However, this opens the door to pragmatic creativity. Given the widespread use of RANS closure for engineering analysis and for design in a wide range of industries, and given its central role in commercial CFD codes, closure modeling will remain a very attractive area of research.

Many experiments and high-fidelity computer simulations have a stated objective of providing data for modeling. However, large data sets have led to few advances in the way that data are used: Model constants are tuned by eyeballing the fit to data, models are assessed by compute and compare, and discrepancies are noted but not explained any better with big data sets than was done with small ones. There is a need for innovative research into data-driven modeling.

New directions in closure modeling per se will likely be toward expanding the phenomena they incorporate, as illustrated by the ideas reviewed herein for incorporating transition and streamline curvature. Inaccuracies near separation lines and inaccurate reattachment lengths, etc., are long-standing motives for new research. Transition models are evolving and should be improved. There are other phenomena (e.g., effects of surface roughness, wall proximity, shock–turbulence interactions, applications to multiphase flow or plasma dynamics) that support either improving the empiricism or examining new formulations.

Hybrid methods invite further questions: Where is LES more reliable than RANS, and where is it less so? Do various blending, shielding, and interpolation functions plausibly select the most suitable mode? On coarse grids, LES can be grid dependent and hence user dependent; from an industrial perspective, robustness is another concern. Can hybrid methods be made resistant to vagaries of gridding? More accurate Reynolds stress closure models can lead to improvements in hybrid simulation. Should the underlying closures be devised specifically for hybrid modeling? These and other issues deserve attention.

**FUTURE ISSUES**

1. Improving predictive accuracy remains a driving force in closure modeling. In many industrial applications, accuracy of turbulence models remains the pacing item in CFD analysis.

2. Schemes to incorporate new phenomena into closure models need to be developed. The demand is growing with the breadth of applications.

3. Various routes of laminar to turbulent transition need to be modeled. The challenge is to predict the location of the transition. Precursors may be Tollmien–Schlichting waves, Klebanoff modes, cross-flow instability, Mack modes at high Mach number, streaks caused by surface roughness, stratified and buoyant instability, etc.

4. Hybrid blending functions should select which of RANS or LES is more reliable within regions of a given flow, with a given grid, with a given numerical scheme, and with a given closure model. Is the ratio of grid size to turbulent scale the right metric? Adaptive methods should continue to be developed.

5. Can methods of data-driven science and engineering lead to improved closure models? Comprehensive data fields can be used to test modeled terms or to extract information on turbulence variables. Methods of machine learning are being explored for correlating model term corrections with invariant parameters.
DISCLOSURE STATEMENT

The author is not aware of any biases that might be perceived as affecting the objectivity of this review.

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