

Turbulent Transport in Single-Phase Flow

*Peter Bernard,
University of Maryland*

Assume that our goal is to compute mean flow statistics such as

$$\overline{U}_i \quad \text{and} \quad K = \frac{1}{2} \overline{u_i^2} \quad \text{where} \quad U_i = \overline{U}_i + u_i$$

One can either:

Pursue DNS (i.e. the "honest" approach) of averaging solutions of the NS eqn:

$$\rho \left(\frac{\partial U_i}{\partial t} + U_j \frac{\partial U_i}{\partial x_j} \right) = - \frac{\partial P}{\partial x_i} + \mu \nabla^2 U_i$$

or pursue RANS (i.e. the "dishonest" approach) of solving the averaged NS eqn:

$$\rho \left(\frac{\partial \overline{U}_i}{\partial t} + \overline{U}_j \frac{\partial \overline{U}_i}{\partial x_j} \right) = - \frac{\partial \overline{P}}{\partial x_i} + \mu \nabla^2 \overline{U}_i - \frac{\partial (\rho \overline{u_i u_j})}{\partial x_j}$$

where $\sigma_{ij}^T \equiv -\rho \overline{u_i u_j} = -\rho R_{ij}$ the Reynolds stress tensor, is modeled.

DNS: Highly accurate but of limited practical usefulness.

RANS: Inaccurate, unreliable, requires empirical modeling, but of widespread use.

LES, a third approach has conceptual problems - though these are usually ignored. In particular, the average of the filtered velocity:

$$\langle U_i \rangle(\mathbf{x}, t) = \int_{\mathfrak{R}^3} G^*(\mathbf{x} - \mathbf{y}) U_i(\mathbf{y}, t) d\mathbf{y}$$

does not necessarily equal the mean velocity, i.e.

$$\overline{U_i} \neq \langle U_i \rangle$$

Moreover, if $u_i = u_i^r + u_i^u$ where $u_i^r = \langle U_i \rangle - \overline{U_i}$

is the resolved part of the velocity fluctuation, then

$$K = \frac{1}{2} \overline{(u_i^r)^2} + \overline{u_i^r u_i^u} + \frac{1}{2} \overline{(u_i^u)^2}$$

Conundrum: if the subgrid energy is large, then K cannot be found. If the subgrid energy is small, then LES is a DNS.

Our interest here is in the RANS approach.

There are 2 basic options:

Direct models for $R_{ij} = \overline{u_i u_j}$

$$R_{ij} = ???$$

or model the R_{ij} equation:

$$\frac{\partial R_{ij}}{\partial t} + \overline{U}_k \frac{\partial R_{ij}}{\partial x_k} = -R_{ik} \frac{\partial \overline{U}_j}{\partial x_k} - R_{jk} \frac{\partial \overline{U}_i}{\partial x_k} - \epsilon_{ij} - \frac{\partial \beta_{ijk}}{\partial x_k} + \Pi_{ij} + \nu \nabla^2 R_{ij}$$

$$\beta_{ijk} \equiv \overline{u_i u_j u_k} + \frac{1}{\rho} \overline{p u_i} \delta_{jk} + \frac{1}{\rho} \overline{p u_j} \delta_{ik}$$

$$\Pi_{ij} \equiv \frac{1}{\rho} \overline{p \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right)}$$

Direct models are most popular and we consider just this case.

The Reynolds stress R_{ij} has a physical interpretation as the flux of the i th component of momentum in the j th direction caused by the fluctuating velocity field.

For non-dense gases the stress tensor in the Navier-Stokes equation has a similar interpretation as representing the flux of the i th component of momentum in the j th direction due to molecular motion.

In the molecular case:

$$C_i = \langle C_i \rangle + c_i \quad \langle C_i \rangle = U_i$$

and the stress tensor is:

$$\sigma_{ij} = -\rho \langle c_i c_j \rangle$$

$$-\rho \langle c_i c_j \rangle = -p \delta_{ij} + \mu \left(\frac{\partial U_i}{\partial x_j} + \frac{\partial U_j}{\partial x_i} \right)$$

Can a similar model for the Reynolds stress tensor be justified?

$$U_i = \overline{U}_i + u_i$$

$$\sigma_{ij}^T \equiv -\rho \overline{u_i u_j} = -\rho R_{ij}$$

$$-\rho R_{ij} = -\frac{2}{3} \rho K \delta_{ij} + \mu_t \left(\frac{\partial \overline{U}_i}{\partial x_j} + \frac{\partial \overline{U}_j}{\partial x_i} \right)$$

**There are very strong reasons for wanting such a model to be true.
In this case the mean momentum equation becomes:**

$$\frac{\partial \bar{U}_i}{\partial t} + \bar{U}_j \frac{\partial \bar{U}_i}{\partial x_j} = -\frac{\partial}{\partial x_i} \left[\frac{\bar{P}}{\rho} + \frac{2}{3} K \right] + \frac{\partial}{\partial x_j} \left[(\nu + \nu_t) \left(\frac{\partial \bar{U}_i}{\partial x_j} + \frac{\partial \bar{U}_j}{\partial x_i} \right) \right]$$

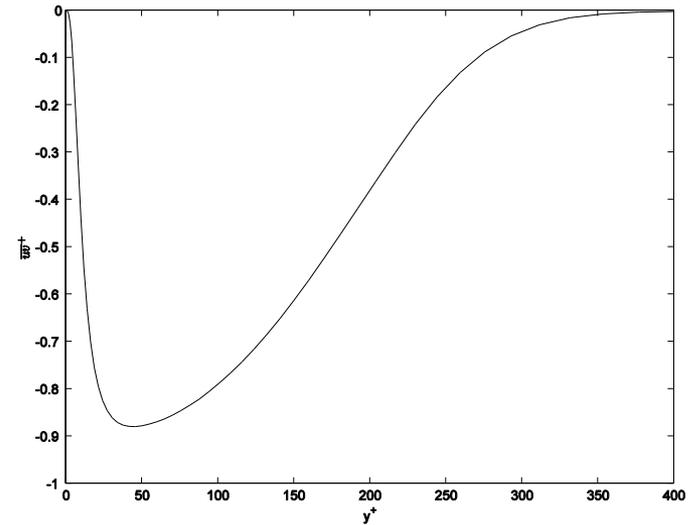
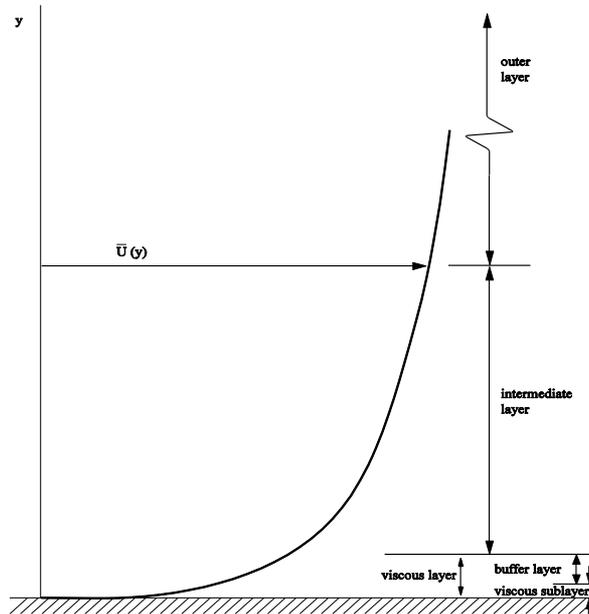
This approach is:

- **easy to install within a NS solver**
 - **relatively well behaved**
 - **relatively inexpensive to solve**

Consider the validity of the molecular transport analogy in the context of a turbulent transport in a unidirectional mean flow such as in a channel or boundary layer:

$$\overline{u'v'}(y) > 0$$

$$\frac{d\overline{U}}{dy}(y) > 0$$

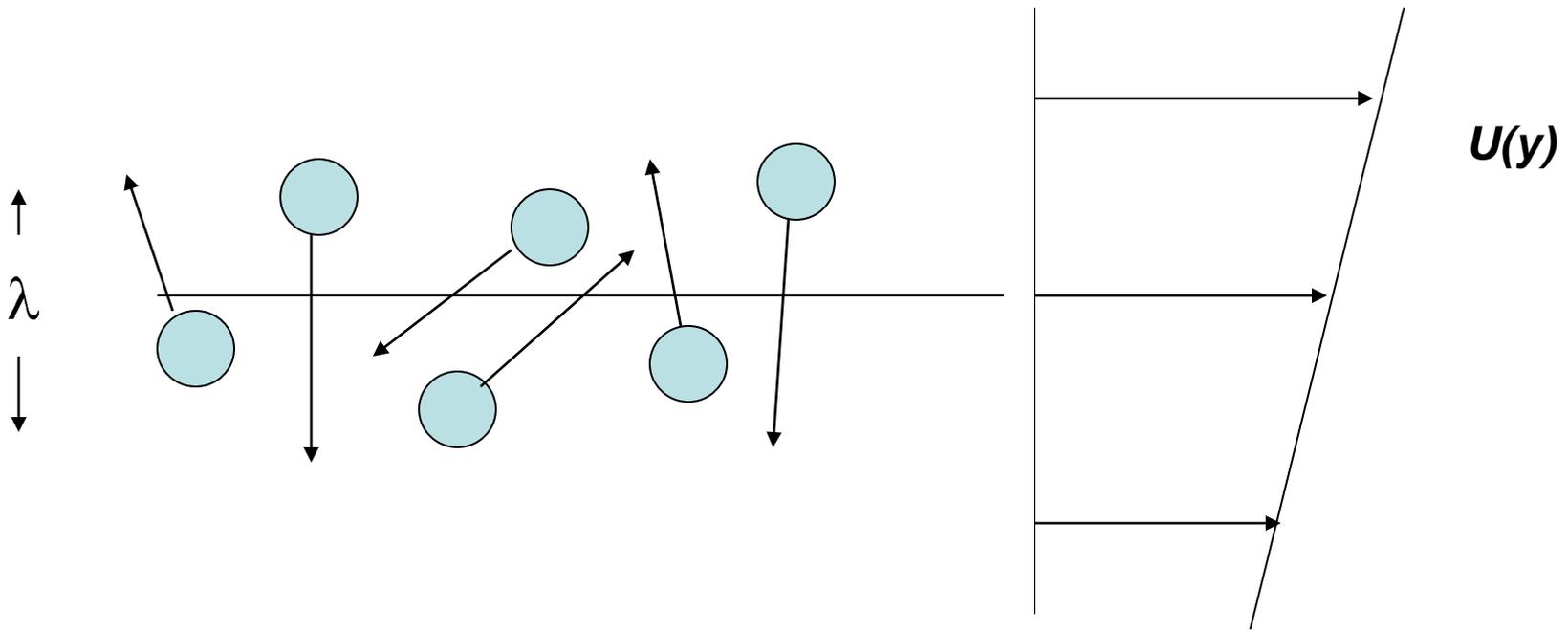


$$\overline{u'v'}(y) < 0$$

In this case:

$$-\rho \langle c_1 c_2 \rangle = \mu \frac{dU}{dy}$$

$$-\rho \overline{u'v'} = \mu_t \frac{d\overline{U}}{dy}$$



Molecules transport momentum, unchanged, over the mean free path, λ , before colliding with other molecules and exchanging momentum. $U(y)$ is linear over λ .

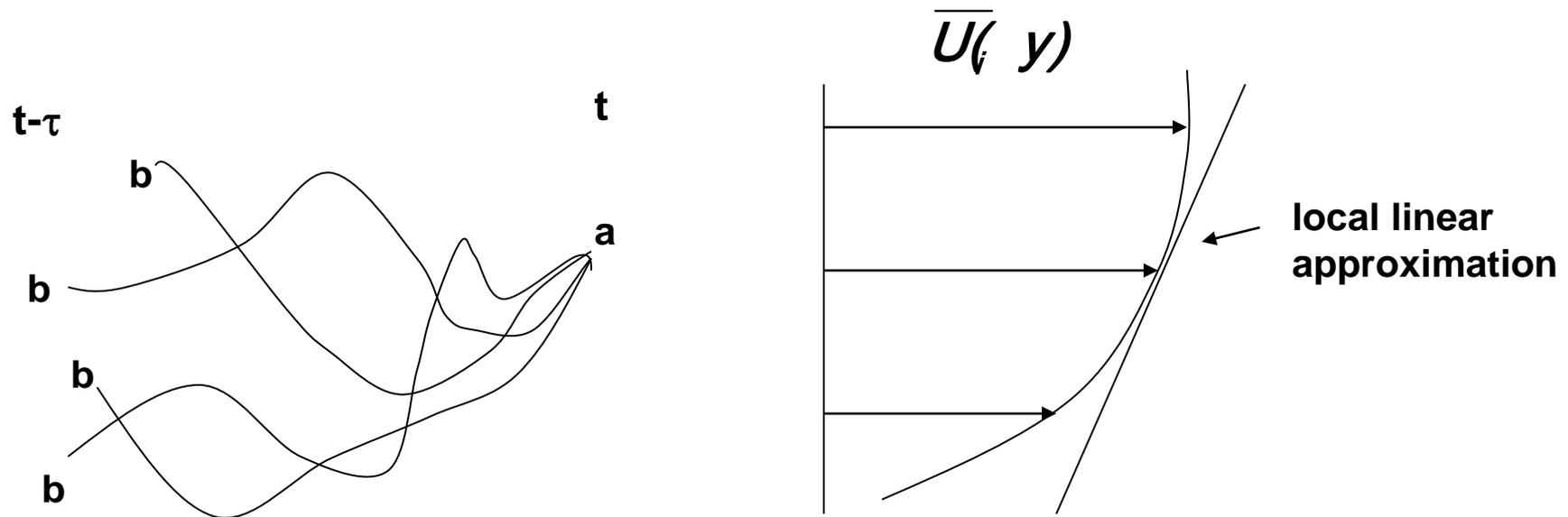
$$\rho \langle c_1 c_2 \rangle \approx \frac{1}{4} \rho c (U(y_0 - \alpha \lambda) - U(y_0 + \alpha \lambda))$$

$$\rho \langle c_1 c_2 \rangle \approx -\mu \frac{dU}{dy}$$

$$\mu = \frac{1}{2} \alpha \lambda \rho c,$$

here: $c = \sqrt{c_i^2}$

To analyze the physical mechanisms behind turbulent transport consider the set of fluid elements that arrive at a given point a at time t .



Unlike the molecular case:

- momentum is not preserved on paths until mixing.
- the idea of "mixing" is undefined
- no obvious separation of scales

Use backward particle paths to evaluate an exact Lagrangian decomposition of the Reynolds shear stress that exposes the underlying physics.

$$\overline{uv} = \overline{u^b v} + \overline{v(\overline{U}^b - \overline{U})} + \overline{v(U - U^b)}$$

goes to 0 as τ increases
(establishes a mixing time).

Thus

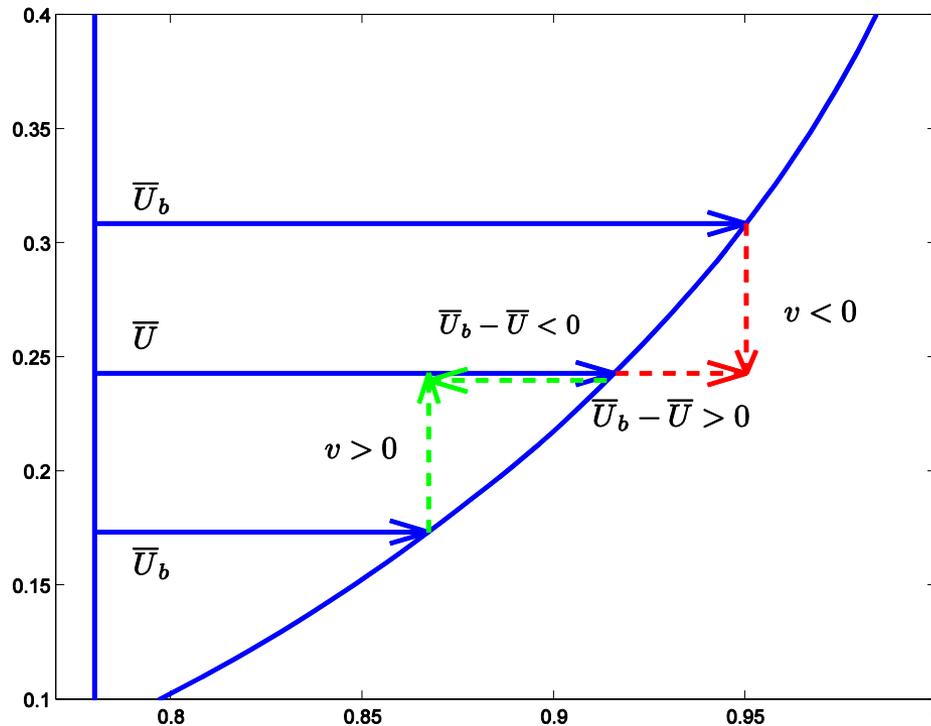
$$\overline{uv} = \overline{v(\overline{U}^b - \overline{U})} + \overline{v(U - U^b)}$$

transport caused by fluid particles carrying, unchanged, the mean momentum at point b to point a.

transport associated with changes in velocity (accelerations) along particle paths.

$$\overline{v \bar{U}^b - \bar{U}}$$

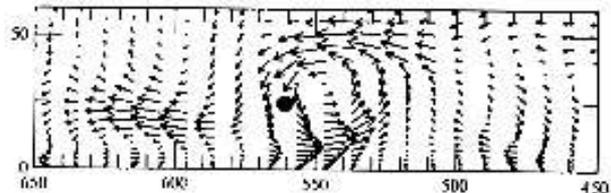
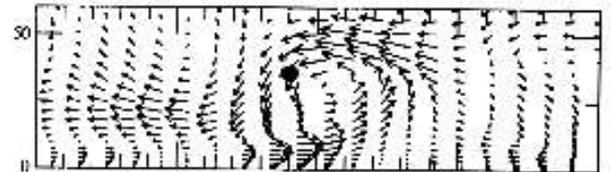
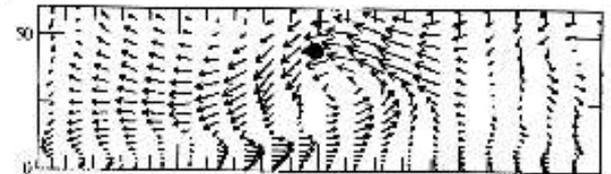
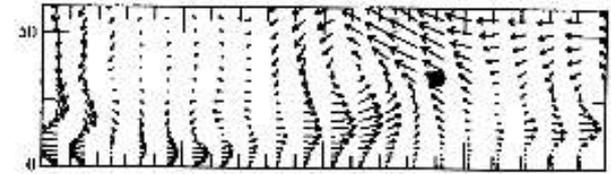
The correlation is created by fluid particle movements within a spatially varying mean field: when $v > 0$ the difference in mean velocity along the path is negative and vice versa.



$$v \bar{U}^b - \bar{U} < 0$$

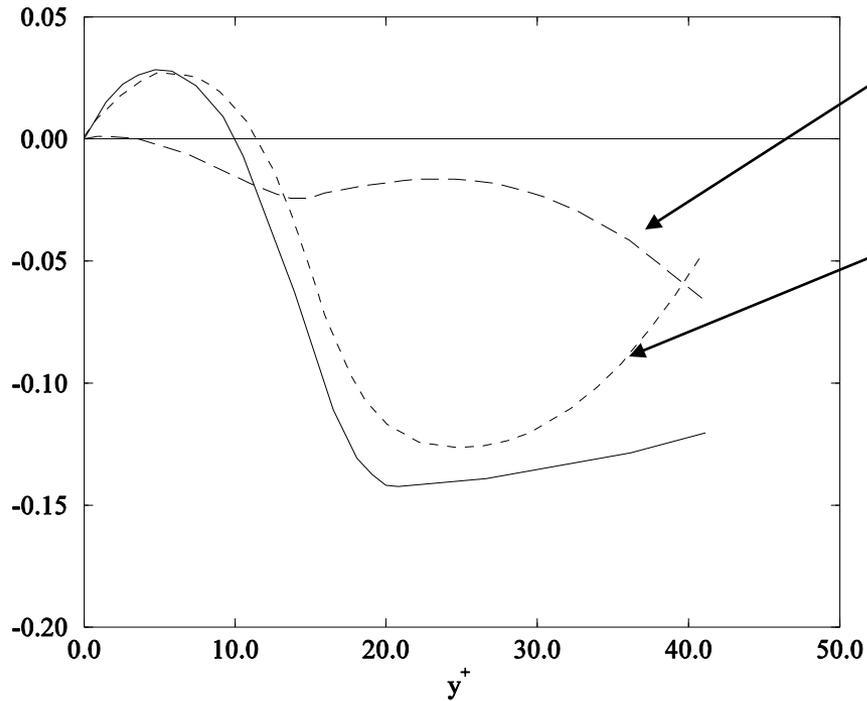
Acceleration transport originates largely in the effect of vortical structures in accelerating fluid particles as they move toward the wall (sweeps) or retarding fluid particles as they eject from near the wall.

Close to the surface, viscous effects retard fast moving fluid particles leading to a decrease in Reynolds stress.

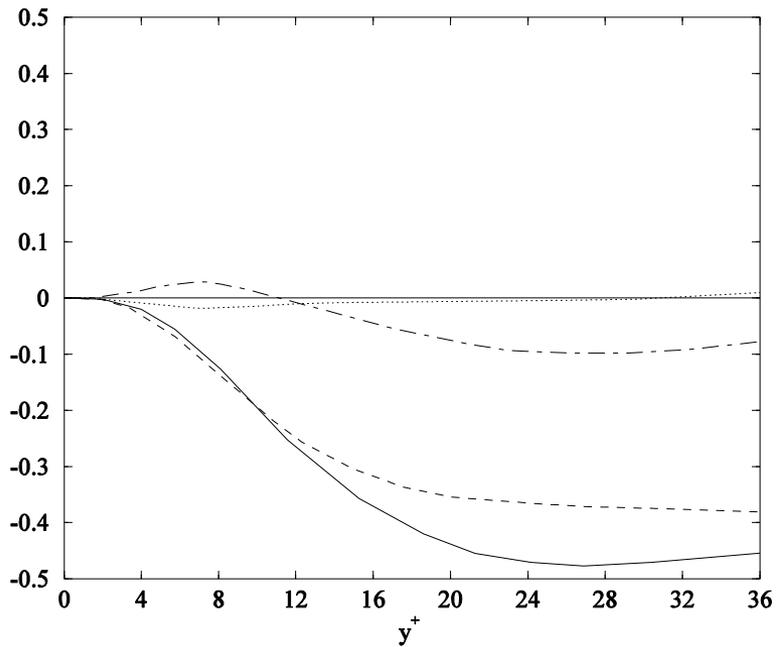


Decomposition of acceleration transport into viscous and pressure effects.

$$\overline{v(U - U^b)} = \int_{t-\tau}^t \overline{v \frac{\partial P}{\partial x}}(s) ds + \int_{t-\tau}^t \overline{v \nabla^2 U}(s) ds$$



Evaluation of the Lagrangian decomposition in channel flow yields:



transport due to
particle accelerations

transport due to
particle displacements

$$\overline{uv} = \overline{v(\overline{U}^b - \overline{U})} + \overline{v(U - U^b)}$$

The Lagrangian analysis can yield a quantitative estimate of the potential errors in a gradient model of the Reynolds stress.

$$\overline{U^b} = \overline{U} - L_2 \frac{d\overline{U}}{dy} + \frac{L_2^2}{2} \frac{d^2\overline{U}}{dy^2} + \dots$$

$$\mathbf{L} \equiv \mathbf{x} - \mathbf{b} = \int_{t-\tau}^t \mathbf{U}(\mathbf{X}(s), s) ds$$

(Mixing length - distance traveled during the mixing time)

$$\overline{v(\overline{U^b} - \overline{U})} = -\overline{v}L_2 \frac{d\overline{U}}{dy} + \frac{\overline{v}L_2^2}{2} \frac{d^2\overline{U}}{dy^2} + \dots$$

gradient term

effect of non-linearity of the mean velocity over the mixing length

An exact decomposition of the turbulent shear stress:

$$\overline{uv} = -T_{22} \overline{v^2} \frac{d\overline{U}}{dy} + \Phi_{12} + \overline{v(U - U^b)}$$

Correct gradient
contribution

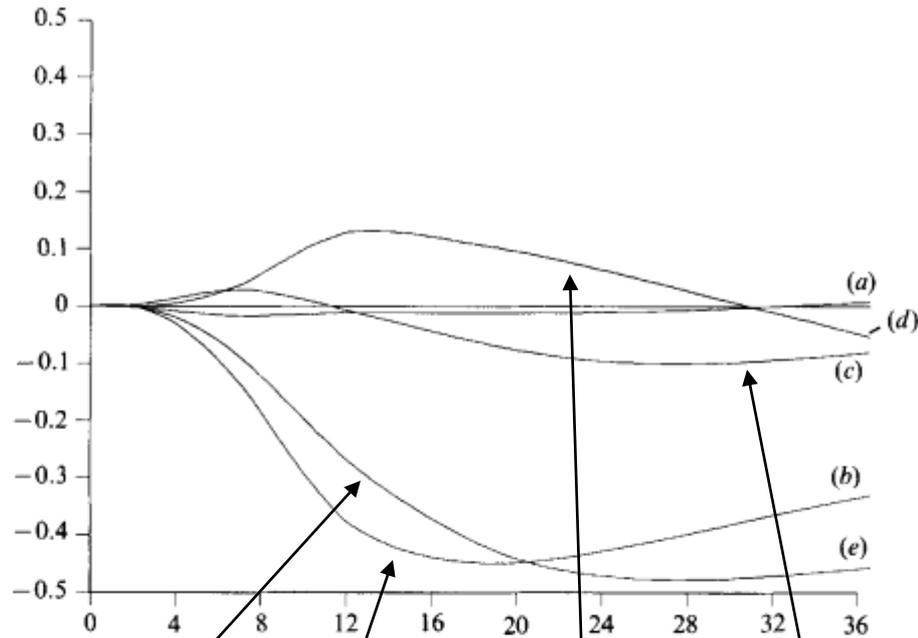
Non-linearity
of mean velocity

Acceleration
effects

where $\overline{vL_2} = \int_{t-\tau}^t \overline{v(t) v(s)} ds = T_{22} \overline{v^2}$

Lagrangian integral time scale

Errors in the gradient model:



$$\overline{uv} = -T_{22} \overline{v^2} \frac{d\overline{U}}{dy} + \Phi_{12} + \overline{v(U - U^b)}$$

Clearly , significant errors are present.

RANS models attempt to compensate for errors by a judicious choice of the eddy viscosity.

Dissatisfaction with linear transport models has fueled interest in models that are non-linear in:

$$S_{ij} = \frac{1}{2} \left(\frac{\partial \bar{U}_i}{\partial x_j} + \frac{\partial \bar{U}_j}{\partial x_i} \right) \quad \text{(rate of strain tensor)}$$

$$W_{ij} = \frac{1}{2} \left(\frac{\partial \bar{U}_i}{\partial x_j} - \frac{\partial \bar{U}_j}{\partial x_i} \right) \quad \text{(rotation tensor)}$$

A typical example of a non-linear model (e.g. Algebraic RS Models):

$$\begin{aligned} R_{ij} = & \frac{2}{3} K \delta_{ij} - 2C_\mu \frac{K^2}{\epsilon} \bar{S}_{ij} + A_1 \frac{K^3}{\epsilon^2} (\bar{S}_{ik} \bar{S}_{jk} - \frac{1}{3} \bar{S}_{kl} \bar{S}_{kl} \delta_{ij}) \\ & + A_2 \frac{K^3}{\epsilon^2} (\bar{W}_{ik} \bar{W}_{jk} - \frac{1}{3} \bar{W}_{kl} \bar{W}_{kl} \delta_{ij}) \\ & + A_3 \frac{K^3}{\epsilon^2} (\bar{S}_{ik} \bar{W}_{jk} + \bar{S}_{jk} \bar{W}_{ik}) + A_4 \frac{K^3}{\epsilon^2} \left(\frac{\partial \bar{S}_{ij}}{\partial t} + \bar{U}_k \frac{\partial \bar{S}_{ij}}{\partial x_k} \right) \end{aligned}$$

Sometimes non-linear models are derived by simplification of RSE models.

Assuming some legitimacy for linear RS models - what is ν_t ?

For molecular transport:

$$\nu = \frac{\mu}{\rho} = \frac{1}{2} \alpha \lambda c \propto UL$$

suggesting that the eddy viscosity depends on the product of velocity (U) and length (L) scales. RANS models vary depending on the choice for U and L.

The K - ϵ closure assumes

$$U \propto \sqrt{K} \quad L \propto K^{3/2} / \epsilon \quad (\text{eddy turnover time})$$

Thus:

$$\nu_t = C_\mu K^2 / \epsilon$$

K equation modeling

$$\frac{\partial K}{\partial t} + \bar{U}_j \frac{\partial K}{\partial x_j} = -\frac{\partial \bar{U}_i}{\partial x_j} R_{ij} - \epsilon - \frac{1}{\rho} \frac{\partial \overline{\rho u_i}}{\partial x_i} + \nu \nabla^2 K - \frac{\partial \overline{u_j (u_i^2/2)}}{\partial x_j}$$

Production

From ϵ equation

$$\frac{\partial}{\partial x_i} \left[\frac{\nu_t}{\sigma_K} \frac{\partial K}{\partial x_i} \right]$$

σ_K is a turbulent
Prandtl number

ε equation

$$\frac{D\epsilon}{Dt} = P_\epsilon^1 + P_\epsilon^2 + P_\epsilon^3 + P_\epsilon^4 + \Pi_\epsilon + T_\epsilon + D_\epsilon - \Upsilon_\epsilon$$

$$P_\epsilon^1 = -2\nu \overline{\frac{\partial u_j}{\partial x_i} \frac{\partial u_j}{\partial x_k} \frac{\partial \bar{U}_i}{\partial x_k}}$$

$$P_\epsilon^2 = -\epsilon_{ij} \frac{\partial \bar{U}_i}{\partial x_j}$$

$$P_\epsilon^3 = -2\nu \overline{u_k \frac{\partial u_i}{\partial x_j} \frac{\partial^2 \bar{U}_i}{\partial x_k \partial x_j}}$$

$$P_\epsilon^4 = -2\nu \overline{\frac{\partial u_i}{\partial x_k} \frac{\partial u_i}{\partial x_j} \frac{\partial u_k}{\partial x_j}}$$

$$\Pi_\epsilon = -2\nu \frac{\partial}{\partial x_i} \left(\frac{\partial p}{\partial x_j} \frac{\partial u_i}{\partial x_j} \right)$$

$$T_\epsilon = -\nu \frac{\partial}{\partial x_k} \left(u_k \frac{\partial u_i}{\partial x_j} \frac{\partial u_i}{\partial x_j} \right)$$

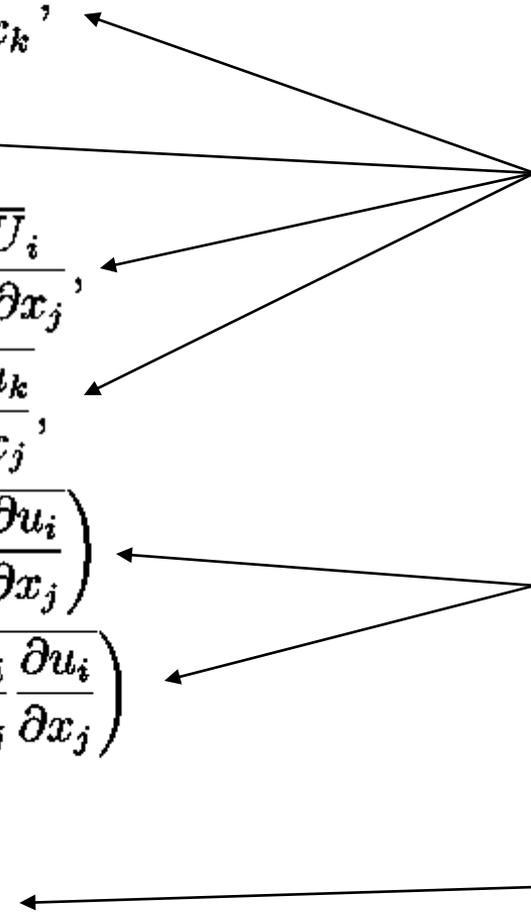
$$D_\epsilon = \nu \nabla^2 \epsilon$$

$$\Upsilon_\epsilon = \nu^2 \overline{\left(\frac{\partial^2 u_i}{\partial x_j \partial x_k} \right)^2}$$

Production

Transport/Diffusion

Dissipation



**Modeling of the ε equation is done in stages
by considering its properties in simplified settings:**

1. isotropic decay.

2. homogeneous shear flow.

3. constant stress layer near solid walls.

The exact equations governing the decay of Isotropic Turbulence:

$$\frac{dK}{dt} = -\epsilon$$

$$\frac{d\epsilon}{dt} = P_\epsilon^4 - \Upsilon_\epsilon = -2\nu \overline{\frac{\partial u_i}{\partial x_l} \frac{\partial u_i}{\partial x_j} \frac{\partial u_l}{\partial x_j}} - 2\nu^2 \overline{\left(\frac{\partial^2 u_i}{\partial x_j \partial x_l} \right)^2}$$

vortex stretching dissipation

After defining:

$$S_K \equiv - \frac{\overline{\left(\frac{\partial u}{\partial x} \right)^3}}{\overline{\left(\frac{\partial u}{\partial x} \right)^2}^{3/2}}$$

(skewness)

$$G = \frac{\overline{\left(\frac{\partial^2 u}{\partial x^2} \right)^2}}{\overline{\left(\frac{\partial u}{\partial x} \right)^2}^2}$$

(palenstrophy)

$$R_T = \frac{K^2}{\nu \epsilon}$$

Reynolds number

$$\frac{dK}{dt} = -\epsilon$$

These may be simplified to:

$$\frac{d\epsilon}{dt} = S_K^* R_T^{1/2} \frac{\epsilon^2}{K} - G^* \frac{\epsilon^2}{K}$$

In the case of self-similarity, e.g.: $f(r, t) = \overline{u \theta u r} = \tilde{f} \left(\frac{r}{\lambda t} \right)$

S_K and G are constant and the system of equations is closed and solvable.

$$\frac{dK}{dt} = -\epsilon$$

$$\frac{d\epsilon}{dt} = S_K^* R_T^{\frac{1}{2}} \frac{\epsilon^2}{K} - G^* \frac{\epsilon^2}{K}$$

Two equilibria exist:

Low R_T : vortex stretching negligible: $\frac{d\epsilon}{dt} = -\frac{7}{5} \frac{\epsilon^2}{K} \quad K \approx t^{-5/2}$

High R_T : vortex stretching and dissipation equilibrate $\frac{d\epsilon}{dt} = -2 \frac{\epsilon^2}{K} \quad K \approx t^{-1}$

In traditional modeling vortex stretching is eliminated creating an opportunity to match decay rate with experiments:

$$\frac{d\epsilon}{dt} = -C_{\epsilon_2} \frac{\epsilon^2}{K} \quad K \approx t^{-1/C_{\epsilon_2}}$$

Homogeneous shear flow $S \equiv \frac{d\bar{U}}{dy}$ is constant everywhere

$$\frac{dK}{dt} = P - \epsilon$$

$$\frac{d\epsilon}{dt} = P_\epsilon^1 + P_\epsilon^2 + P_\epsilon^4 - \Upsilon_\epsilon$$

$$P_\epsilon^1 + P_\epsilon^2 = -2\nu \overline{\omega_1 \omega_2} S$$

exact equations

Assume: $P = -\overline{uv} \frac{d\bar{U}}{dy} = C_\mu \frac{K^2}{\epsilon} S^2$

$$\frac{\overline{\omega_1 \omega_2}}{\zeta} \sim \frac{\overline{uv}}{2K}$$

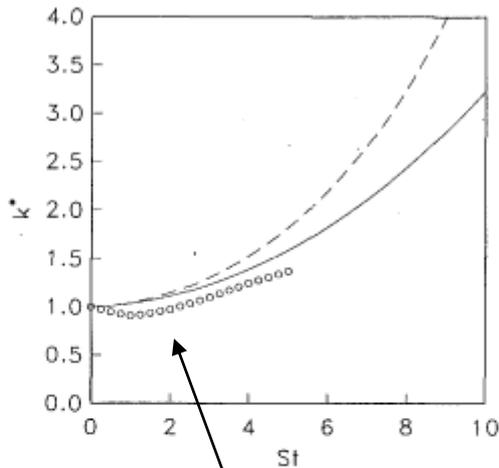
$\zeta \equiv \overline{\omega_i^2} \longrightarrow$ enstrophy

$$P_\epsilon^1 + P_\epsilon^2 = C_{\epsilon 1} C_\mu K S^2$$

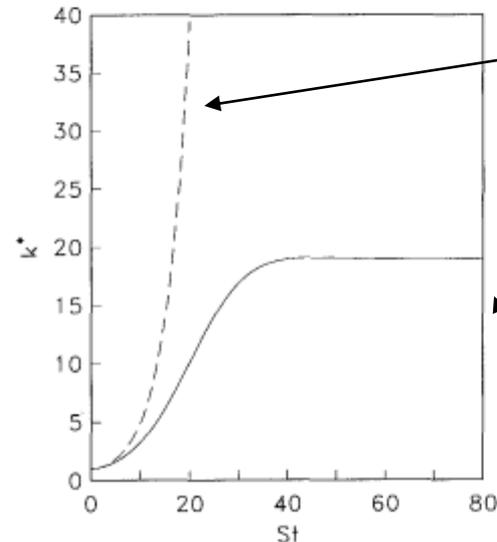
Modeled Equations for Homogeneous Shear Flow

$$\frac{dK}{dt} = C_{\mu} \frac{K^2}{\epsilon} S^2 - \epsilon$$

$$\frac{d\epsilon}{dt} = C_{\epsilon_1} C_{\mu} K S^2 + C_{\epsilon_3} R_T^{\frac{1}{2}} \frac{\epsilon^2}{K} - C_{\epsilon_2} \frac{\epsilon^2}{K}$$



C_{ϵ_1} chosen to match experiments



Without vortex stretching: blow up.

With vortex stretching: prod = diss equilibrium

ϵ equation modeling

homogeneous shear flow model -
calibrated to give correct K growth

$$C_{\epsilon_1} \frac{\epsilon}{K} \mathcal{P}$$

$$\frac{\partial}{\partial x_i} \left[\left(\frac{\nu_t}{\sigma_\epsilon} \right) \frac{\partial \epsilon}{\partial x_i} \right]$$

$$\frac{D\epsilon}{Dt} = P_\epsilon^1 + P_\epsilon^2 + P_\epsilon^3 + P_\epsilon^4 + \Pi_\epsilon + T_\epsilon + D_\epsilon - \Upsilon_\epsilon$$

$$-C_{\epsilon_2} \frac{\epsilon^2}{K}$$

Isotropic turbulence model - calibrated
to give a decay rate consistent with data

K-ε Closure (high Re form)

$$\frac{\partial \bar{U}_i}{\partial t} + \bar{U}_j \frac{\partial \bar{U}_i}{\partial x_j} = -\frac{\partial}{\partial x_i} \left[\frac{\bar{P}}{\rho} + \frac{2}{3} K \right] + \frac{\partial}{\partial x_j} \left[\left(\nu + \frac{C_\mu K^2}{\epsilon} \right) \left(\frac{\partial \bar{U}_i}{\partial x_j} + \frac{\partial \bar{U}_j}{\partial x_i} \right) \right]$$

$$\frac{\partial K}{\partial t} + \bar{U}_j \frac{\partial K}{\partial x_j} = \frac{C_\mu K^2}{\epsilon} \left(\frac{\partial \bar{U}_i}{\partial x_j} + \frac{\partial \bar{U}_j}{\partial x_i} \right) \frac{\partial \bar{U}_i}{\partial x_j} - \epsilon + \frac{\partial}{\partial x_i} \left[\left(\nu + \frac{C_\mu K^2}{\sigma_K \epsilon} \right) \frac{\partial K}{\partial x_i} \right]$$

$$\frac{\partial \epsilon}{\partial t} + \bar{U}_j \frac{\partial \epsilon}{\partial x_j} = C_{\epsilon_1} C_\mu K \left(\frac{\partial \bar{U}_i}{\partial x_j} + \frac{\partial \bar{U}_j}{\partial x_i} \right) \frac{\partial \bar{U}_i}{\partial x_j} - C_{\epsilon_2} \frac{\epsilon^2}{K} + \frac{\partial}{\partial x_i} \left[\left(\nu + \frac{C_\mu K^2}{\sigma_\epsilon \epsilon} \right) \frac{\partial \epsilon}{\partial x_i} \right]$$

$$C_\mu = 0.09 \quad C_{\epsilon_1} = 1.44 \quad C_{\epsilon_2} = 1.92$$

$$\sigma_K = 1.0 \quad \sigma_\epsilon = 1.3$$

Calibration of the K-ε Closure

In the "constant stress layer"

$$\overline{uv} \approx -U_\tau^2$$

Assume:

$$\mathcal{P} = -\overline{uv} \frac{d\overline{U}}{dy} \approx \epsilon$$

and the model:

$$\overline{uv} \approx -C_\mu \frac{K^2}{\epsilon} \frac{d\overline{U}}{dy}$$

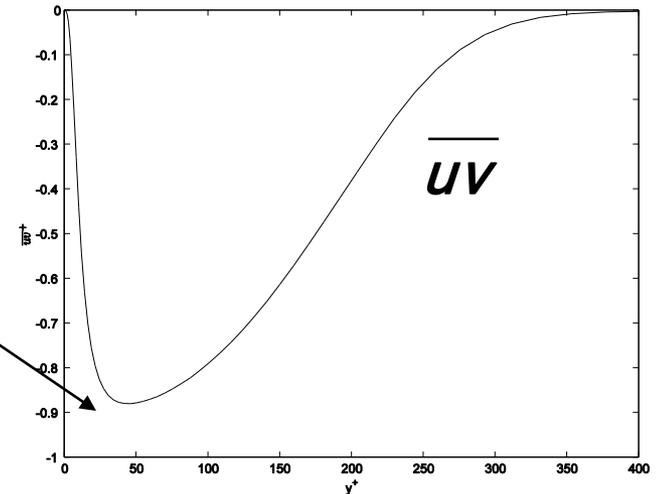
Then:

$$K = \frac{U_\tau^2}{\sqrt{C_\mu}}$$

Moreover, if $\overline{U}^+(y^+) = \frac{1}{\kappa} \ln(y^+) + B$ then: $\epsilon = \frac{U_\tau^3}{\kappa y}$

Substituting these results into the ε equation gives:

$$\frac{\sqrt{C_\mu} \sigma_\epsilon}{\kappa^2} (C_{\epsilon_2} - C_{\epsilon_1}) = 1$$



Near-wall modeling

Boundary conditions:

$$K(0) = 0 \quad \varepsilon(0) = 2\nu \frac{\partial^2 K}{\partial y^2}(0) = 2\nu \left(\frac{\partial \sqrt{K}}{\partial y}(0) \right)^2$$

Among the problems with the high Re modeling near a boundary:

$$v_t = C_\mu K^2 / \varepsilon \rightarrow T_{22} \overline{v^2}$$

Introduce a wall function to force the equivalence:

$$v_t = C_\mu f_\mu K^2 / \varepsilon$$

In effect, $f_\mu K \rightarrow \overline{v^2}$

Other problems that have to be fixed near a wall:

$$\frac{\partial \epsilon}{\partial t} + \bar{U}_j \frac{\partial \epsilon}{\partial x_j} = C_{\epsilon_1} C_{\mu} K \left(\frac{\partial \bar{U}_i}{\partial x_j} + \frac{\partial \bar{U}_j}{\partial x_i} \right) \frac{\partial \bar{U}_i}{\partial x_j} - C_{\epsilon_2} \frac{\epsilon^2}{K} + \frac{\partial}{\partial x_i} \left[\left(\nu + \frac{C_{\mu} K^2}{\sigma_{\epsilon} \epsilon} \right) \frac{\partial \epsilon}{\partial x_i} \right]$$

**K → 0 at wall so
dissipation blows up**

At the wall surface: $\Pi_{\epsilon}(0) + D_{\epsilon}(0) = \Upsilon_{\epsilon}(0)$

yet no explicit model for Π_{ϵ} has been assumed in high Re model.

Low Re model for the ϵ equation near walls.

$$\frac{\partial \epsilon}{\partial t} + \bar{U}_j \frac{\partial \epsilon}{\partial x_j} = C_{\epsilon_1} f_1 \frac{\epsilon}{K} \mathcal{P}_K - C_{\epsilon_2} f_2 \frac{\epsilon \tilde{\epsilon}}{K} + \frac{\partial}{\partial x_i} \left[\left(\nu + \frac{\nu_t}{\sigma_\epsilon} \right) \frac{\partial \epsilon}{\partial x_i} \right] + F$$

here (e.g.) $\tilde{\epsilon} = \epsilon - 2u \left(\frac{\partial \sqrt{K}}{\partial y} (0) \right)^2$

wall functions

$$\nu_t = C_\mu f_\mu K^2 / \epsilon$$

What to expect from the popular RANS models:

- 1. The predictions of RANS models in their standard form, can be both acceptable or unacceptable depending on the desired accuracy, naivety of the user and other factors.**
- 2. It is very common to make ad hoc changes to the values of constants and even to add additional modeling expressions in order to improve accuracy, or to force the solution to acquire desired physical attributes. The idea is that some aspect of physics is lacking in the original model that needs to be compensated for.**
- 3. Changing the properties of models can bring the solutions closer to one set of data and further from another set of data.**
- 4. Sometimes model alterations - with no basis in physics - are made as a last resort to force better results: e.g. "clipping"**
- 5. RANS solutions sometimes are regarded as successful if only one part of the solution is captured - the part that is of interest.**

6. Adding additional physics to RANS calculations can be especially difficult - two layers of inaccuracy: the underlying turbulence and the new physical model. Different models of the physics (e.g. particle dispersion, chemistry, combustion) can react differently to the same underlying RANS modeling.

7. A numerical calculation with a RANS scheme may converge for one set of input parameters and not converge for a similar case of the same flow.

8. The quality of one particular RANS model may appear to be better than it is because it performs better than other models.

9. Very often computational speed is considered more important than accuracy.

10. In some flows, complaints about steady RANS solutions have led to the use of URANS (Unsteady RANS) in which features such as vortex shedding are considered to be part of the mean (albeit transient) field.

11. Many research studies have compared LES predictions to RANS predictions. Sometimes RANS is as good as LES, sometimes LES is better, sometimes the added accuracy of LES is not justified by the cost.

12. RANS is increasingly being used to model the wall region of LES since the local DNS resolution that one would hope for is often not feasible.