

NORDITA/FLOW Spring School on Turbulent Boundary Layers

Lecture VII

Issues in DNS

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Overview

(following “*Primer on DNS of turbulence*”)

- **Introduction***
 - Context, seminal studies*
- **The Reynolds-number constraint***
- **Numerical considerations***
 - **Spatial discretisation***
 - Spectral methods*
 - Finite-difference methods*
 - (Temporal discretisation)
 - (Boundary/initial conditions)
 - (Code validation & resolution guidelines)
- (Outlook)

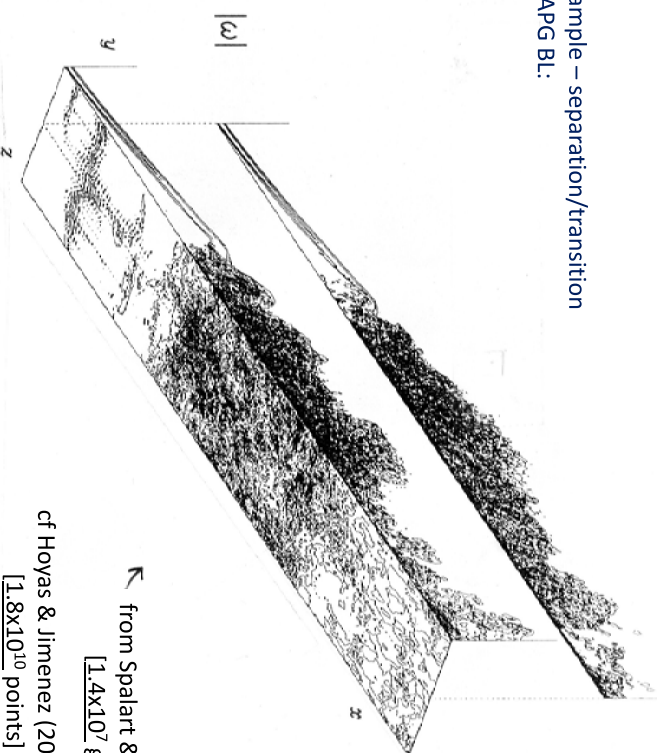
- **Case study: Near-wall similarity and DNS* (Lecture VIIb)**

* Covered here

Pioneers

- Orszag & Patterson (1972) – *Homogeneous isotropic turbulence* [Fourier³ spectral]
- Kim, Moin & Moser (1987) – $R_\tau=180$ *turbulent plane channel/Poiseuille flow* [Fourier²/Chebyshev- τ spectral]
- Spalart (1988) – $R_\theta=1410$ *ZPG boundary layer* [Fourier²/Jacobi spectral]

Example – separation/transition
in APG BL:



↖ from Spalart & Strelets (2000)
[1.4×10^7 grid points]

cf Hoyas & Jimenez (2006):
[1.8×10^{10} points]

Direct numerical simulation (DNS)

Conceptually straightforward:

- Discretise 3D Navier-Stokes equations on fine grid
- Time step resolving the fine-scale motion (*unsteady* questions must be solved)
- Collect statistics, etc just as for a short-duration experimental realisation of flow:

DNS = ‘*numerical experiment*’

- Resolve all spatial & temporal scale of motion

Or LES...



DNS is not RANS...



DNS = ‘Exact’ high-fidelity CFD



Benefits:

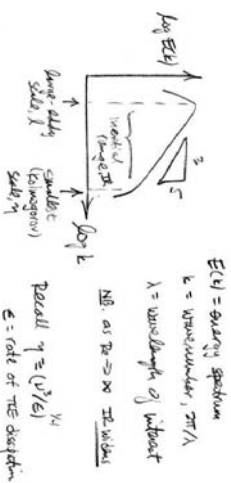
- Capable of providing complete knowledge of turbulent flow
- Complete control of IC/BCs and parameters & equations (can "play with nature"!)
 - $\therefore$ *ideal research tool*

Drawbacks:

- Computationally very expensive, due to need to resolve fine-scale motion: **HPC required**
- Computational costs scale horribly with Reynolds number...

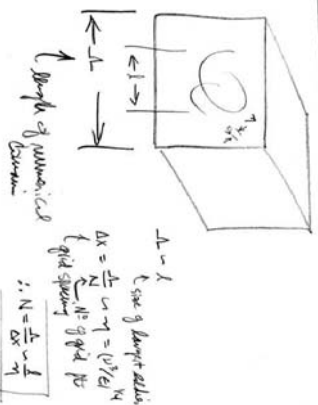
The Reynolds-number constraint:

Range of Scales in turbulent flow:



DNS case with 3 homogeneous directions, (x, y, z) :

DNS domain:



Assume LS & SS in equilibrium (e.g. all scales have adjusted to present state of mean (order)

$\Rightarrow \epsilon = \text{rate at which TKE cascades from LS}^*$ to be dissipated by ν at SS, and

that $\lambda \sim \ell^{3/2}$ where $\ell \sim \text{TKE}^{1/2}$ is

velocity scale of large eddies

$$\text{where } N \sim \frac{L}{\eta} \sim \left(\frac{L}{\ell}\right)^{3/2} \left(\frac{\ell}{\eta}\right)^{3/2} = \frac{L^{3/2}}{\eta^{3/2}}$$

for turbulent flow $\ell^{3/2}$ for LS/SS equil case

$$\text{we define } Re_T \equiv \frac{L}{\eta} \sim \frac{L}{\ell} \Rightarrow \frac{L^{3/2}}{\eta^{3/2}} \sim N$$

← 湍流雷诺数: Re_T

NB: In general $Re_T = Re_T(x, t)$ (ie in a time-avg 3D field variable), but since here $\frac{\partial \text{mean}}{\partial x} = 0$ (due to 3D homogeneity) a single Re_T at each time gives static flow.

$$\text{since } N \sim \frac{L^{3/2}}{\eta^{3/2}} \sim Re_T^{3/2}$$

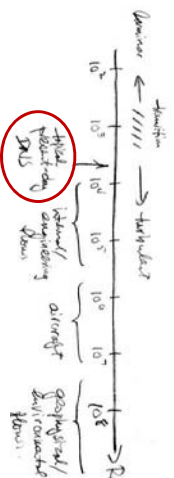
For complete DNS: 'cost' $\sim N^3 \cdot N_t$

← time step Δt (if $\Delta t \sim \eta$)
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$$\text{where } N_t \sim \frac{T_{\text{sim}}}{\Delta t} \sim \frac{1}{\Delta t} \sim N$$

... and $\boxed{\text{lost in } Re^3} \leftarrow \text{guess!}$
 Δ Turn to increase Re by 10,
 lost goes up by 1000! (1 day $\rightarrow$ 3 years!)

Different $\Lambda=N(Re)$ relationships for other types of flows,
 but in general cost $\sim Re^n$ where $n \sim 2-3$. Consequently...

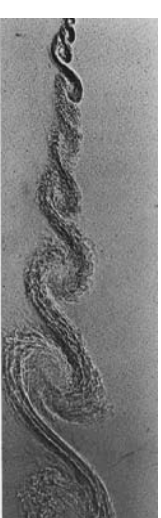


► Most engineering flows are beyond the reach of
 DNS – by a large factor (it is *not* a design tool)

BUT:

- DNS can give complete and accurate knowledge of turbulence at moderate Reynolds numbers, and is therefore *ideally suited for studying fundamental physics*, testing RANS/SGS models
- And results at lower Re can often be safely extrapolated to higher Re via *Reynolds-number similarity* (i.e. Re-independence of Large Scales)...

Re ‘similarity’ ...

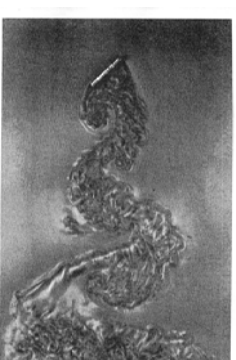


126. Large-scale structure in a turbulent mixing layer. Nitrogen flows at 1000 cm/sec issuing from a nozzle, impinging on a flat plate. The flow is visualized by a smoke sheet. The large-scale structure is characterized by the presence of a dominant scale. Their spacing at the downstream side of the layer is larger than near the beginning. Photograph by H. Kanaoka, N.A.S. Conf. Ser. of Tokyo, 1976.

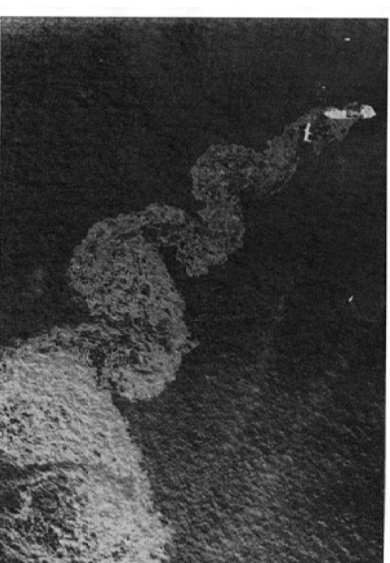


127. Coherent structure at higher Reynolds numbers. This flow is an above view at twice the pressure. Doubling the Reynolds number has produced more stable scale ratio. M. R. Kanaoka, N.A.S. Conf. Ser. of Tokyo, 1976. Photo by H. Kanaoka, 1979.

(from *An Album of Fluid Motion*, van Dyke 1982)



127. Wake of an inclined flat plate. The wake behind a plate at 45° angle of attack to turbulent at a Reynolds number of 4000. The flow is visualized by a smoke sheet. The large-scale structure is characterized by the presence of a dominant scale. Photograph by H. Kanaoka, N.A.S. Conf. Ser. of Tokyo, 1976. Photo by H. Kanaoka, 1979.



125. Wake of a grounded tankship. The tanker *Argo* is shown in the photograph. The wake is characterized by the presence of a dominant scale. The large-scale structure is characterized by the presence of a dominant scale. Photograph by H. Kanaoka, N.A.S. Conf. Ser. of Tokyo, 1976. Photo by H. Kanaoka, 1979.

(from van Dyke 1982)

...Real-world relevance of DNS

- For phenomena and statistics associated with *largest scales* (mixing, transport), DNS at “high enough Re” captures typical high-Re behaviour
- And for wall-bounded turbulence (for which effective $Re \rightarrow 0$ as $y^+ \rightarrow 0$) DNS accurately represents at least *some* of the critical near-wall dynamics (streaks, etc) of high-Re flows
- For some cases (μ AVs, very high Ma) DNS *can* reach design conditions

Numerical methods for DNS

- ▶ Must provide accurate realisation of the flow over wide range or length/time scales, plus accurate time history of the flow, to yield correct turbulence statistics and flow structures
 - for *spatial discretisation*, traditional choice is **spectral methods**... (see ‘*Primer*’ for discussion of *finite-difference methods* and *temporal* discretisation)

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 - **Spatial discretisation**
 - Spectral methods
 - **Finite-difference methods**
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- (Outlook)
- **Case study: Near-wall similarity and DNS (*Lecture VIIb*)**

Spatial discretization options:

- Standard (low-order) finite-difference (FD)/finite volume (FV) methods -
 - + Allow complex geometries: *gridding flexibility*
 - Accuracy often unacceptable (especially for *upwind schemes*)
- Spectral methods – [ideal for simple incompressible flows]
 - + Optimal accuracy-to-gridpoint ratio
 - Limited to simple geometries: difficulties with discontinuities (e.g. shocks)
- High-order FD methods (e.g. Pade schemes) – [popular for computational aeroacoustics, compressible turbulence]
 - + Good compromise between accuracy and flexibility
 - Coding more involved
- Finite-element/spectral-element methods – [emerging techniques for complex DNS: riblets, roughness elements, etc]
 - + Arbitrary geometries allowed via unstructured grids
 - Unstructured grids lead to inefficient solution algorithms
- Viscous vortex methods – [current computational research topic]
 - + Can accommodate high-Re bluff-body flows
 - Extension to 3D not straightforward

Spectral methods

Definition:

Spectral methods $\equiv$ spatial discretization via linear combination (expansion) of global basis functions that are orthogonal eigenfunctions of an appropriate Sturm-Liouville problem.

$$Q(x) = \sum_{n=0}^N a_n \underbrace{\phi_n(x)}_{\substack{\text{expansion coeffs.} \\ \text{S-L eigen functions}}} \quad \begin{array}{l} \text{dep variable} \\ (p, u, T, \dots) \end{array}$$

linear combination of smooth functions, each of which span entire flow domain

Sturm-Liouville ODE:

$$\frac{d}{dx} \left[p(x) \frac{d\phi_n}{dx} \right] + \left[\lambda_n w(x) - q(x) \right] \phi_n(x) = 0$$

$\uparrow$ eigenvalue
 $\uparrow$ weight function (over $a \leq x \leq b$)

where λ_n are countable (discrete) set of eigenvalues

$\lambda_1, \lambda_2, \dots$

and for each λ_n there is an associated eigenfunction ϕ_n

$\phi_1, \phi_2, \dots$ which satisfies

orthogonality relation:

$$\int_a^b \phi_n(x) \phi_m(x) w(x) dx = \alpha_n \delta_{nm},$$

Examples:

$\bullet p=1, w=1, d=0 \Rightarrow \phi_n = e^{ik_n x}$, $k_n = \frac{2\pi n}{L}$, $L=b-a$
 $\uparrow$ Fourier Series

$\bullet p=\sqrt{1-x^2}, w=\frac{1}{\sqrt{1-x^2}}, d=0 \Rightarrow \phi_n = T_n(x)$
 $\uparrow$ Chebyshev polynomials

$\bullet p=1-x^2, w=1, d=0 \Rightarrow \phi_n = L_n(x)$
 $\uparrow$ Legendre polynomials

• etc...

Given $Q(x)$, use orthogonality to find a_n :

$$Q(x) = \sum_{n=0}^N a_n \phi_n(x)$$

$$\Rightarrow w(x) \phi_m(x) Q(x) = \sum_{n=0}^N a_n w(x) \phi_m(x) \phi_n(x)$$

$$\Rightarrow \int_a^b w(\tau) \phi_m(\tau) Q(\tau) d\tau = \sum_{n=0}^N a_n \underbrace{\int_a^b w(\tau) \phi_m(\tau) \phi_n(\tau) d\tau}_{\alpha_n \delta_{nm}}$$

$$= \sum_{n=0}^N a_n \alpha_n \delta_{nm} = a_m \alpha_m$$

$$\Rightarrow \boxed{a_m = \frac{1}{\alpha_m} \int_a^b w(\tau) \phi_m(\tau) Q(\tau) d\tau}$$

$m=0, 1, \dots, N$

Spectral DNS:

- $u(x, t) \leftarrow u_N(x, t) = \sum_{n=0}^N \underset{\substack{\uparrow \\ \text{time}}}{a_n(t)} \underset{\substack{\downarrow \\ \text{space}}}{\phi_n(x)}$
- Into N-S eqn $\mathcal{L}(u) = 0$, yields non-zero Residual $R_N = \mathcal{L}(u_N) \leftarrow (\text{error})$

- Minimise error by setting weighted integral of R_N to zero:

$$\int_0^D u(x) \phi_m(x) R_N dx = 0 \text{ for } m=0, \dots, N$$

$\Downarrow$

$$[A] \frac{d\vec{a}}{dt} = [B] \vec{a} + \vec{f}$$

$$[A] \frac{d\vec{a}}{dt} = [B] \vec{a} + \vec{f}$$

- $(N+1) \times (N+1)$ matrices
 elements defined analytically
 (via orthogonality)
- Apply time-marching algorithm (Runge-Kutta, Crank-Nicholson, ...) to $(N+1)^2$ system to solve for $\vec{a} = [a_0, a_1, \dots, a_N]$ at new time

$\nwarrow$ include nonlinear convective term
 $\therefore$ pseudo-spectral method (Orszag 1971)

- Transform $\vec{a}$ to physical space $\hat{u}(x)$ and compute nonlinear/convective terms
- Transform nonlinear terms to wave space, form RHS of $(N+1)^2$ system
- Repeat...

> Features of spectral methods

- If basis-function variations correctly 'fit' dependent-variable variations (as does Fourier series for statistically homogeneous directions), then:
 - the expansion coefficients a_n go to zero as $N \rightarrow \infty$,

and - they do so such that discretisation error decreases faster than any (!) power of $1/N$.

(apply integration by parts repeatedly to

$$a_n = \frac{1}{\alpha_n} \int_a^b Q(x) W(x) \phi_n(x) dx$$

if

$$\lim_{N \rightarrow \infty} (N^{m+1} a_N) \rightarrow 0, \text{ where } m \text{ is order of highest derivative of } Q(x)$$

$\therefore$ If Q is k th differentiable ($m=k$), then $a_N \rightarrow 0$ faster than any power p of $1/N$.

$$\left(\text{cf. FDA: error} \sim (\Delta x)^p \sim (1/N)^p \right)$$

$\therefore$ spectral methods have 'exponential' or ' N -order' convergence.

- Verification that $|a_n| \rightarrow 0$ as $n \rightarrow N$ provides 'a posteriori' check that spatial resolution is sufficient (i.e. no grid refinement study needed to ensure spectral DNS results are 'broadly correct')

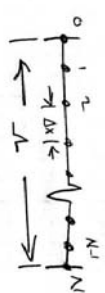
(BUT see Lecture VIIIb Case Study)

▷ Fourier series [the classic/quintessential spectral method]

Fourier interpolant: $u(x) = \sum_{l=0}^{N-1} \hat{u}_l e^{i k_l x}$ where $k_l = \frac{2\pi l}{L}$ and $u(0) = u(L)$

continuous variable, physical space, wave space

At discrete points $x_l = l \Delta x = l L/N$



we have $u(x_l) \equiv u_l = \sum_{l=0}^{N-1} \hat{u}_l e^{i \frac{2\pi l}{N} x_l}$

⇒

$$u_l = \sum_{l=0}^{N-1} \hat{u}_l e^{2\pi i l x_l / N}$$

Discrete Fourier Transf. (DFT)
Backward Transf. (wave → phys. space)

$$\textcircled{1} \Rightarrow \sum_{l=0}^{N-1} u_l e^{-2\pi i n l / N} = \sum_{l=0}^{N-1} \hat{u}_l \sum_{l=0}^{N-1} e^{2\pi i l (l-n) / N}$$

→

Recall Geometric Series: $\sum_{l=0}^{N-1} z^l = \begin{cases} \frac{z^N - 1}{z - 1} & \text{for } |z| \neq 1 \\ N & \text{for } |z| = 1 \end{cases}$

Here we have $z = e^{2\pi i (l-n) / N}$
 $\begin{cases} \neq 1 & l-n \neq pN \\ 1 & l-n = 0 \end{cases}$ where $p = 0, \pm 1, \pm 2, \dots$

$$\sum_{l=0}^{N-1} \left[e^{2\pi i l (l-n) / N} \right]$$

$$= \begin{cases} \frac{e^{2\pi i (l-n)} - 1}{e^{2\pi i (l-n) / N} - 1} = 0 & l-n \neq pN \\ N & l-n = 0 \end{cases}$$

$$= N \delta_{ln}$$

▷ Fast DFT (FFT) Cooley-Tukey (1965)

Let $N = 2^m \Rightarrow N \in [1, 2, 4, 8, 16, 32, 64, \dots]$

$$\begin{aligned}
 u_1 &= \sum_{j=0}^{N-1} u_j e^{2\pi i j / N} = \sum_{\substack{j=0 \\ \text{even}}}^{N-2} \hat{u}_j e^{2\pi i j / N} + \sum_{\substack{j=1 \\ \text{odd}}}^{N-1} \hat{u}_j e^{2\pi i j / N} \\
 &\underbrace{\mathcal{O}(N^2)}_{\text{even}} \quad \quad \quad \underbrace{\mathcal{O}(N/2)}_{\text{odd}} \\
 &= \sum_{n=0}^{N/2-1} \hat{u}_{2n} e^{2\pi i 2n / N} + \sum_{n=0}^{N/2-1} \hat{u}_{2n+1} e^{2\pi i (2n+1) / N} \\
 &\underbrace{\mathcal{O}(N/2)}_{\text{the length}} \Rightarrow \underbrace{\sum_{n=0}^{N/2-1} \hat{u}_{2n} e^{2\pi i 2n / N}}_{\mathcal{O}(N/2)} + e^{2\pi i / N} \underbrace{\sum_{n=0}^{N/2-1} \hat{u}_{2n+1} e^{2\pi i 2n / N}}_{\mathcal{O}(N/2)}
 \end{aligned}$$

$$\text{so that } \langle u \rangle \Rightarrow \sum_{j=0}^{N-1} u_j e^{-2\pi i j / N} = \sum_{j=0}^{N-1} \hat{u}_j N \delta_{jn} = N \hat{u}_n$$

$$\hat{u}_n = \frac{1}{N} \sum_{j=0}^{N-1} u_j e^{-2\pi i j / N} \quad \text{Forward DFT (plug } \Rightarrow \text{ wave)}$$

wrt DFT...

Note:

- (1) and (2) are a Fwrd/Bkwd transform pair
- Forward and Backward DFT are both Matrix-Vector Multiply operations $[M]\{V\}$ with:

$$M_{nj} = (e^{\pm 2\pi i / N})^{nj} \text{ and } v_j = \hat{u}_j \text{ or } u_j$$

- Thus, obtaining the $N \hat{u}_n$'s from the u_j 's (or vice versa) requires $\mathcal{O}(N^2)$ multiply-add operations
- $\therefore$ Very expensive...

Strategy: Further decompose each of smaller DFT's as before until there are N single-pt transforms to perform!

Cost: N (single-pt transform) + a 'resulting' (recombination) of N sets for each splitting.

Since $N = 2^m (\Rightarrow m = \log_2 N)$ there are a total of $\log_2 N$ splittings

$\therefore$ total cost of FFT scales as $N \cdot \log_2 N$

(Makes Fourier-spectral DNS feasible)

▷ Periodicity of FT $\rightarrow$ negative wave numbers

$$u_1 = \sum_{k=0}^{N-1} \hat{u}_k e^{2\pi i k x/N} = \sum_{k=0}^{N/2-1} \hat{u}_k e^{2\pi i k x/N} + \sum_{k=N/2}^{N-1} \hat{u}_k e^{2\pi i k x/N}$$

$$\sum_{p=-N/2}^{-1} \hat{u}_{N+p} e^{2\pi i (p+N) x/N}$$

$$(let p = k - N)$$

$$= \sum_{p=-N/2}^{-1} \hat{u}_{N+p} e^{2\pi i p x/N} = \sum_{p=-N/2}^{-1} \hat{u}_p e^{2\pi i p x/N}$$

Notation
Convention: $\hat{u}_{N+p} \leftarrow \hat{u}_p$ for $-\frac{N}{2} \leq p \leq -1$

$$\therefore u_1 = \sum_{k=0}^{N/2-1} \hat{u}_k e^{2\pi i k x/N} + \sum_{p=-N/2}^{-1} \hat{u}_p e^{2\pi i p x/N}$$

$$\underline{Q1}: u_1 = \sum_{k=-N/2}^{N/2-1} \hat{u}_k e^{2\pi i k x/N}$$

$\leftarrow$ conventional form.

Negative wave numbers
allows straightforward
manipulation between $|k|$
(ie $|k| = \frac{m\lambda}{L}$) and number
of oscillations

NB. The highest wavenumber that can be
captured on a given uniform
grid is $k_{max} = \frac{\pi}{\Delta x}$

(cf. Nyquist frequency from signal processing)

$$\frac{2\pi(N/2)}{\Delta x} = \pi N / \Delta x = \pi / \Delta x$$

And if grid spacing Δx is too coarse
for the spatial variations in $0 \leq x \leq L$
(ie. if k_{max} of the grid is smaller than
the actual max k), then the Fourier
coefficients that are supported by the
grid will be corrupted by Aliasing Errors.

Implication for DNS:

Nonlinear/convective terms in N-S eqns
increase the span of wavenumbers from

$$-\frac{N}{2} \leq n \leq \frac{N}{2} \text{ to } -N \leq n \leq N$$

(span req'd to capture
all dep. var. variations) (span of NL
terms)

$\uparrow$ must be resolved
by numerical scheme too!

Cure:

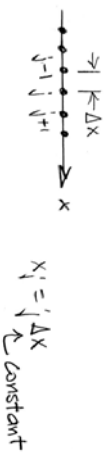
Increase resolution of grid when computing the nonlinear terms; if $1.5 \times N$ collocation/quadrature points are used, results free from aliasing errors (AKA '2-rule').
(For compressible flows, for which (NL) = $2\rho u_i u_j / \Delta x_i$ and PHS of N-S Eqs involve ρ , the situation is much worse, and cannot be exactly cured, by say ($\frac{3}{2}$ -rule).)

Fourier series can be used to analyze other spatial discretizations via modified/effective wavenumber k^*

Example: consider upwind FDA $\frac{du}{dx} = \frac{u_j - u_{j-1}}{\Delta x}$

and use it to approximate derivatives of homogeneous/periodic field on uniform grid Δx

Actual field $u(x) = \sum_k \hat{u}(k) e^{ikx}$



Apply upwind FDA:

$$\begin{aligned} \frac{du}{dx} \Big|_j &= \frac{u_j - u_{j-1}}{\Delta x} = \frac{1}{\Delta x} \sum_k \hat{u}(k) \left[e^{ikx_j} - e^{ikx_{j-1}} \right] \\ &= \frac{1}{\Delta x} \sum_k \hat{u}(k) e^{ikx_j} \left[1 - e^{-ik\Delta x} \right] \end{aligned}$$

$$\therefore \frac{du}{dx} \Big|_j = \sum_k \left(\frac{1}{\Delta x} (1 - e^{-ik\Delta x}) \right) \hat{u}(k) e^{ikx_j}$$

cf. Actual $\frac{du}{dx} \Big|_j = \sum_k ik \hat{u}(k) e^{ikx_j}$

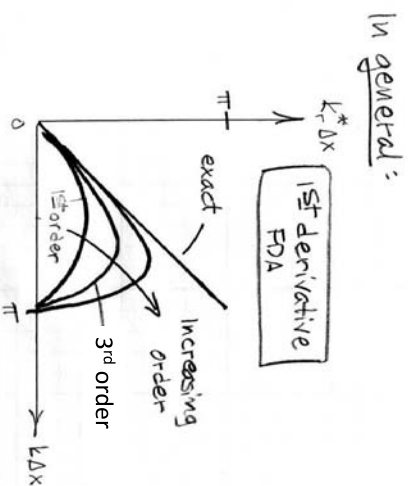
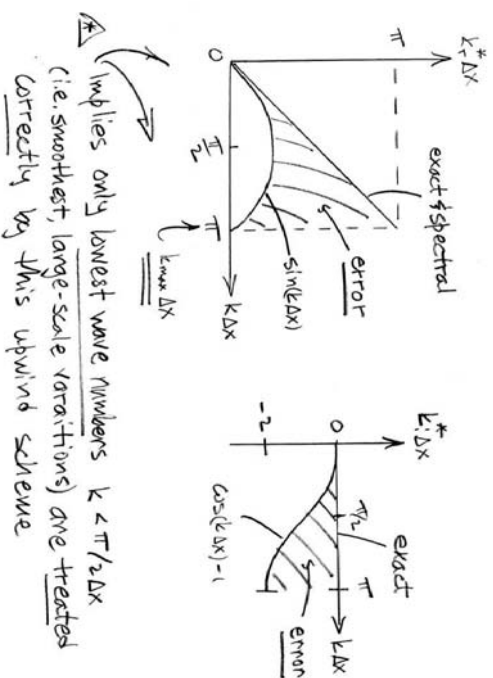
define modified wavenumber k^*

for this FDA:

$$\begin{aligned} ik^* &= \frac{1}{\Delta x} \left[1 - \underbrace{(\cos(k\Delta x) - i \sin(k\Delta x))}_{e^{-ik\Delta x}} \right] \\ &= \frac{1}{\Delta x} [1 - \cos(k\Delta x) + i \sin(k\Delta x)] \end{aligned}$$

thus $k^* \Delta x = \underbrace{\sin(k\Delta x)}_{k^* \Delta x} + i \underbrace{(\cos(k\Delta x) - 1)}_{k^* \Delta x}$

Real part k^* gives dispersion (phase) error
Imag part k^* gives dissipation error



Note: Modified wave number analysis can also be used to quantify errors associated with FDA of other derivatives (e.g. 2nd derivative):

$$\text{exact: } \frac{d^2 u}{dx^2} = -k^2 u$$

$$\text{FDA: } \frac{\hat{\Delta}^2 u}{\Delta x^2} = -k_*^2 u$$

defines modified wave number associated with 2nd derivative FDA used.

Example (cont'd)...

... Apply upwind scheme to 1D wave eqn (linear version of LHS of N-S eqns):

$$\frac{\partial u}{\partial t} + c \frac{\partial u}{\partial x} = 0 \Rightarrow \sum_k \frac{d \hat{u}}{dt} e^{ikx} + \sum_k c i k \hat{u} e^{ikx} = 0$$

$$u(x,t) = \sum_k \hat{u}(k,t) e^{ikx} \Rightarrow \sum_k \frac{d \hat{u}}{dt} e^{ikx} + \sum_k c i k \hat{u} e^{ikx} = 0$$

$$\Rightarrow \frac{d \hat{u}}{dt} + i c k \hat{u} = 0 \Rightarrow \hat{u}(k,t) = \hat{u}(k,0) e^{-i c k t}$$

$$\text{for FDA: } \frac{\partial u}{\partial t} + c \frac{\partial u}{\partial x} = 0 \Rightarrow \hat{u}(k,t) = \hat{u}(k,0) e^{-i c_* k t}$$

Note: This analysis considers only spatial FDA errors - No account taken of errors due to time-marching scheme. (see 'Primer')

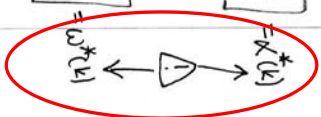
Since $k^* = k_r^* + i k_i^*$ we have

$$\hat{u}(k, t) = \hat{u}(k, 0) e^{-k^* t} = e^{-k_r^* t} e^{-i k_i^* t}$$

with decay rate $\alpha^* = -k_i^* = c \left[1 - \frac{\cos(k \Delta x)}{\Delta x} \right]$

and phase speed $\omega^* = c k_r^* = c \left[\frac{\sin(k \Delta x)}{\Delta x} \right] = \omega^*(k)$

Note that decay rate and phase speed introduced by FDA both depend on wavenumber.



Dispersion/phase error:

exact: $x = ct$

FDA: $x = ct \frac{\sin(k \Delta x)}{k \Delta x} \rightarrow \begin{cases} ct & \text{as } k \rightarrow 0 \text{ (OK)} \\ 0 & \text{as } k \rightarrow \pi/\Delta x \text{ (!)} \end{cases}$

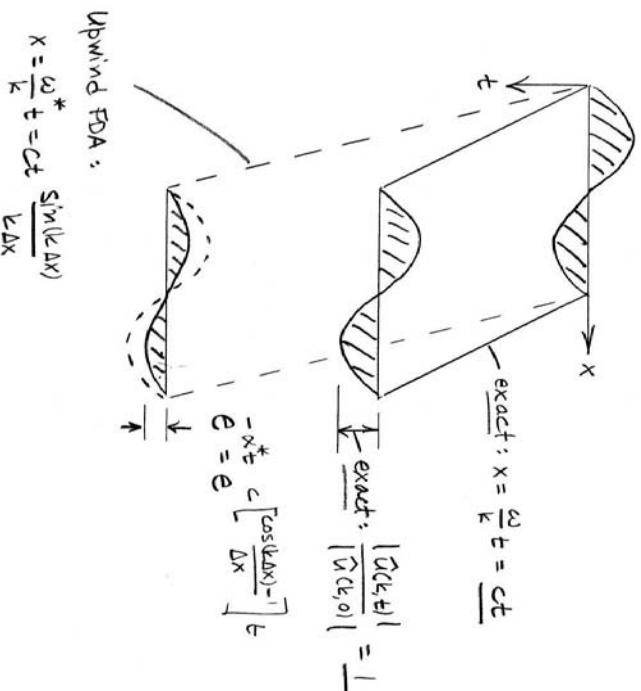
Thus the higher the k (smaller the wavelength, spatial scales) the slower it converts. Dispersion error

Dissipation error:

exact: $e^{-\alpha t} = 1$

FDA: $e^{-\alpha^* t} = e^{-c \left[\frac{\cos(k \Delta x)}{\Delta x} - 1 \right] t} \rightarrow \begin{cases} 1 & \text{as } k \rightarrow 0 \text{ (OK)} \\ 0 & \text{as } k \rightarrow \pi/\Delta x \text{ (!)} \end{cases}$

$\therefore$ the higher the k , the faster it decays. Dissipation error



Upwind FDA:

$$x = \frac{\omega}{k} t = ct \frac{\sin(k \Delta x)}{k \Delta x}$$

Conclusions

- Upwind schemes do not represent the largest resolvable wavenumbers correctly
 - Much finer grid required to match spectral methods
 - Apply to turbulence simulation with caution!
- Do not judge the accuracy/fidelity of a DNS (or *any* CFD solution) solely by the total number of grid points

Near-wall similarity and DNS:

1. **The resilience of the logarithmic law to pressure gradients – Evidence from DNS**
2. **The hunt for κ**

Near-wall similarity and DNS

Lecture VIIb

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– UK Turbulence Consortium –

EPSRC

Engineering and Physical Sciences
Research Council

HPC
CAPABILITY COMPUTING



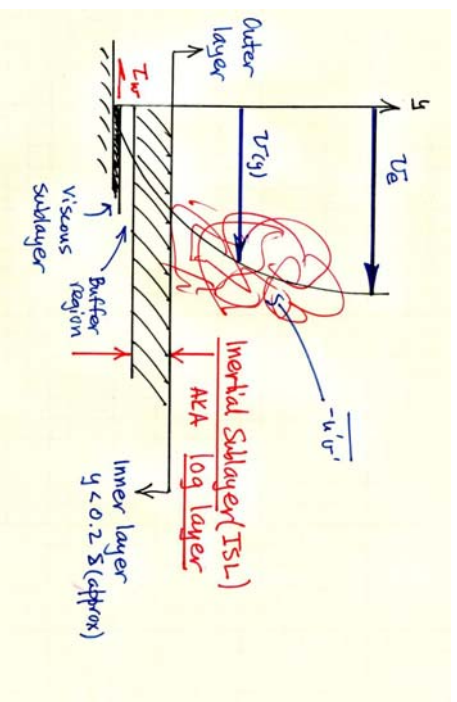
Issues

- How **Robust** and **universal** is the *log law*?...
...Effect of:
 - Flow type/boundary conditions?
 - Reynolds number?
 - Pressure gradients (i.e. $d\tau/dy \neq 0$)?

Approach

- Fully spectral DNS of:
 - Plane Couette-Poiseuille flow
 - Pressure-driven Ekman layer (3DBL)

Review: Near-wall similarity of low-speed zero-pressure-gradient (ZPG) turbulent boundary layers:



ISL Assumptions: $U = U(y, \tau_w, \nu) \xrightarrow{DA} \frac{dU}{dy} = \frac{u_\tau}{y} f(y^+)$

where $u_\tau = \tau_w / \rho$
and $y^+ = y u_\tau / \nu$

As $y^+ \rightarrow$ large enough, effect of ν becomes negligible, and $f(y^+) \rightarrow \frac{1}{K}$, where K is (presumably) a universal constant

$$\therefore \frac{dU}{dy} = \frac{u_\tau}{Ky} \Rightarrow U = \frac{u_\tau}{K} \ln y^+ + C$$

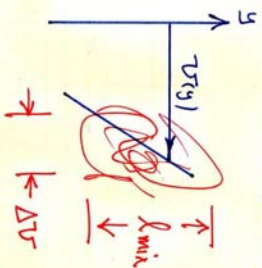
Option (1): $\boxed{\frac{u_\tau}{dU/dy} = Ky}$ "(ISL1)"

Recall Eddy viscosity: $-\overline{u'v'} \equiv \nu_\tau \frac{dU}{dy}$
 $\Rightarrow \nu_\tau = \frac{-\overline{u'v'}}{dU/dy}$

ISL Assumptions: $-\overline{u'v'} = u_\tau^2$
 $dU/dy = u_\tau / Ky \Rightarrow \nu_\tau = u_\tau Ky$

$\rightarrow$ Option (2): $\boxed{\frac{-\overline{u'v'}}{dU/dy} = Ky}$ "(ISL2)"

Algebraic mixing length:



Assume $u'v' \Delta T \equiv l_{mix} \left| \frac{dT}{dy} \right|$

and $v'v' - v'v' = -\Delta T$

$$\Rightarrow -\overline{u'v'} \sim (\Delta T)^2$$

$$\therefore -\overline{u'v'} = l_{mix}^2 \left(\frac{dT}{dy} \right)^2$$

$$\Rightarrow l_{mix} \equiv \frac{\sqrt{-\overline{u'v'}}}{dT/dy}$$

and

$$\nu_T = l_{mix}^2 \frac{dT}{dy}$$

ISL Assumptions: $l_{mix} \sim y$ ($l_{mix} \equiv \sqrt{-\overline{u'v'}}$)

$$\left[\begin{array}{l} -\overline{u'v'} = u_c^2 \\ \frac{dT}{dy} = \frac{u_c}{K y} \end{array} \right] \Rightarrow l_{mix} = \frac{u_c}{u_c/K y} = K y$$

→ Option (3): $\boxed{\frac{\sqrt{-\overline{u'v'}}}{dT/dy} = K y}$ — "ISL3"

Note: $-\overline{u'v'} = u_c^2 \Rightarrow \frac{dT}{dy} = 0$ (i.e. ZPG BL only)

Recap: Near-wall/ISL similarity:

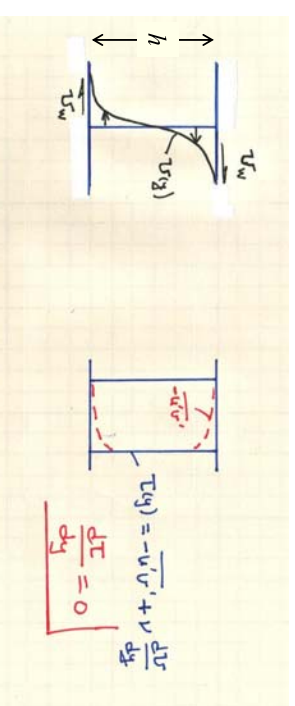
$$\underbrace{\frac{u_c}{dT/dy}}_{\text{log law}} = \underbrace{\frac{-\overline{u'v'}/u_c}{dT/dy}}_{\text{eddy viscosity}} = \underbrace{\frac{\sqrt{-\overline{u'v'}}}{dT/dy}}_{\text{mixing length}} = K y \quad \text{for ZPG}$$

(1)

(2)

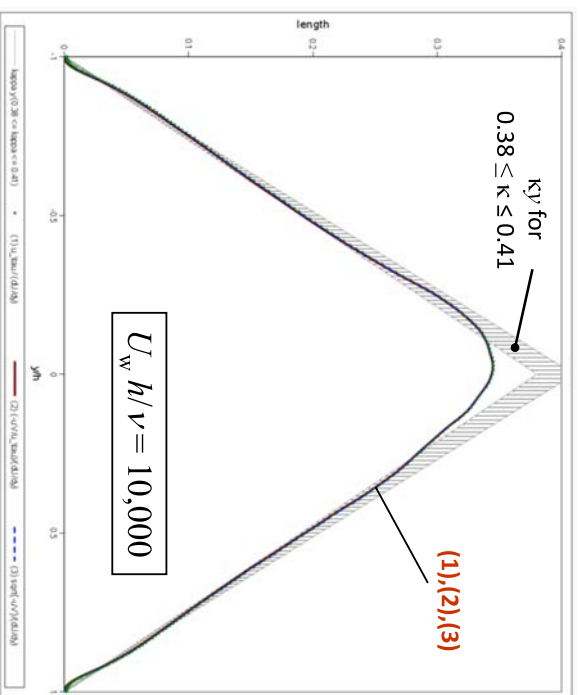
(3)

Compare with DNS of plane turbulent Couette flow:



(Results courtesy of R Johnstone, U Southampton, 2009)

Couette flow DNS ($dP/dx=0$)



(✓)

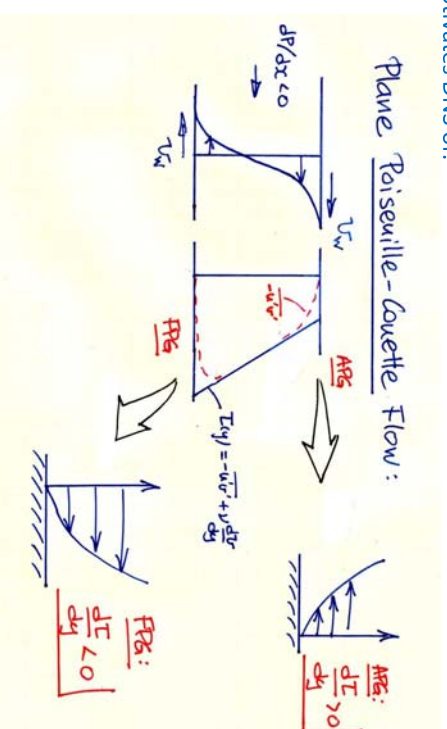
Options:

$$\left. \begin{aligned} (1) \quad \frac{U_\tau}{dy/dy} &\stackrel{?}{=} k_y \\ (2) \quad \frac{-\overline{u'v'}/U_\tau}{dy/dy} &\stackrel{?}{=} k_y \\ (3) \quad \frac{\sqrt{-\overline{u'v'}}}{dy/dy} &\stackrel{?}{=} k_y \end{aligned} \right\} \text{depend on } U_\tau \text{ (i.e. } \tau_w)$$

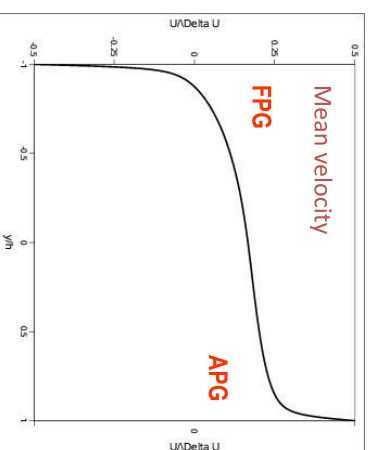
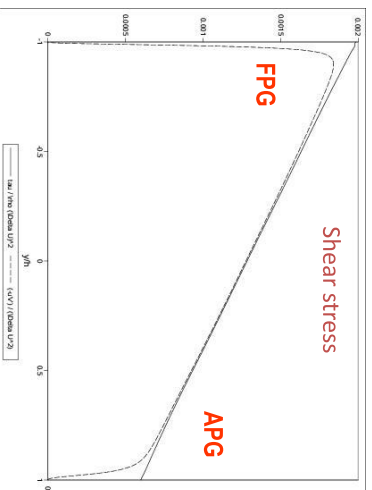
$\leftarrow$ local quantity

Motivates DNS of:

Plane Poiseuille-Couette Flow:



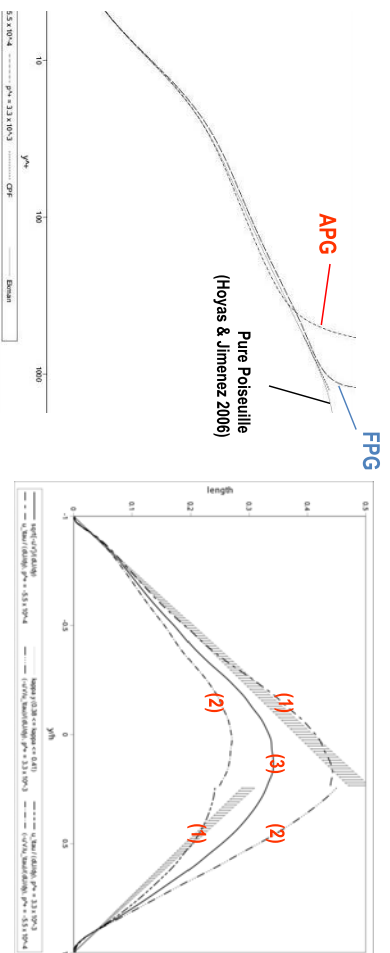
Couette-Poiseuille Flow



Conclusion:

The *log law* is *more resilient to pressure gradients* than eddy-viscosity and mixing length (Johnstone *et al* 2010; cf Galbraith, Sjolander & Head 1977, and Huang & Bradshaw 1995)

Couette-Poiseuille: The Bottom Line



- LHS: Both walls “trying hard” to have log law...
- RHS: Three similarity candidates vs $\kappa y \dots$
 - Theory with $0.38 \leq \kappa \leq 0.41$
 - The Log Law (1) is clear winner
 - Qualitative (rather than sharp, asymptotic) but important result

Recap: Log law characteristics:

Q1. How robust?

A. Quite!

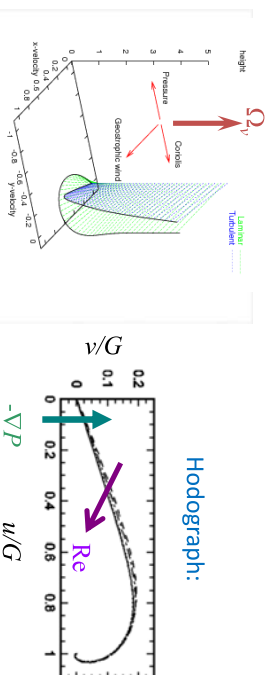
Q2. How universal? (i.e. is κ constant?)

A. TBD...

...What does DNS imply?...

Turbulent (pressure-driven) Ekman layer:

- Balance between pressure gradient, Coriolis and “friction”
⇒ **3D boundary layer...**



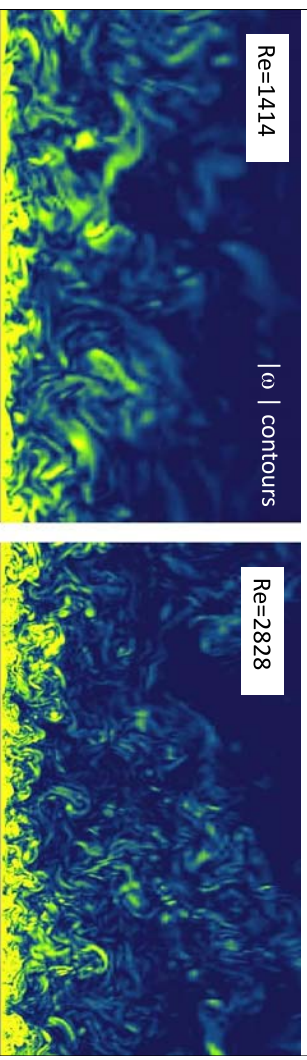
- Defining parameter: **Reynolds number** $Re = GD/\nu$, where
 G ← freestream/geostrophic wind speed
 $D = (2\nu/f)^{1/2}$ ← viscous boundary-layer depth
 $f = 2\Omega_v$ ← Coriolis/rotation parameter
 $\nu = \mu/\rho$ ← kinematic viscosity

Ekman-layer DNS

(Spalart, Coleman & Johnstone 2008, *Phys Fluids*)

- $Re = 1000, 1414, 2000$ and 2828

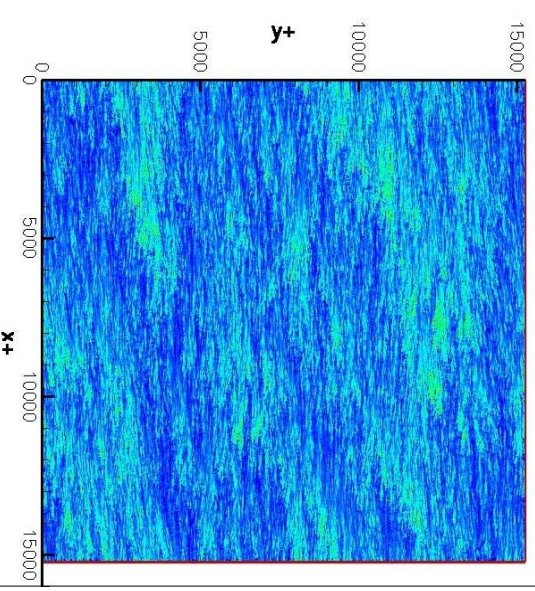
Ekman Layer DNS (Spalart *et al.* 2008)



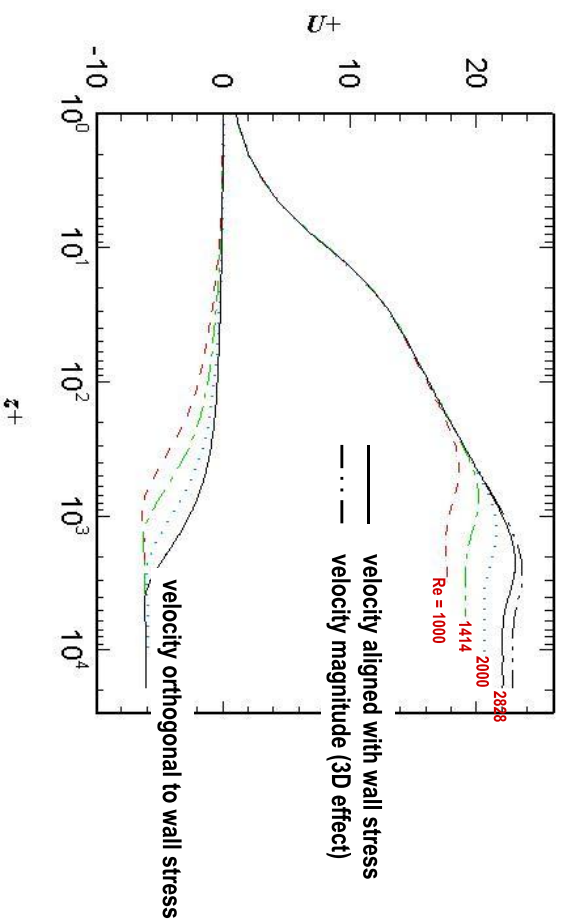
- Relevance for near-wall similarity?...

Ekman-Layer DNS at $Re = 2828$

- Coriolis term allows BL homogeneous in x, y and t
- Pressure gradient, equivalent to channel at $Re_\tau = 1250$
- Boundary-layer thickness $\delta \approx 5000\nu/u_\tau$
- Fully spectral Jacobi/Fourier BL code
- $768 \times 2304 \times 204$ (=360M) quadrature/collocation points
- Patch over $15,000^2$ in wall units, i.e. 150 streaks side-by-side
- Note the “mega-patches”

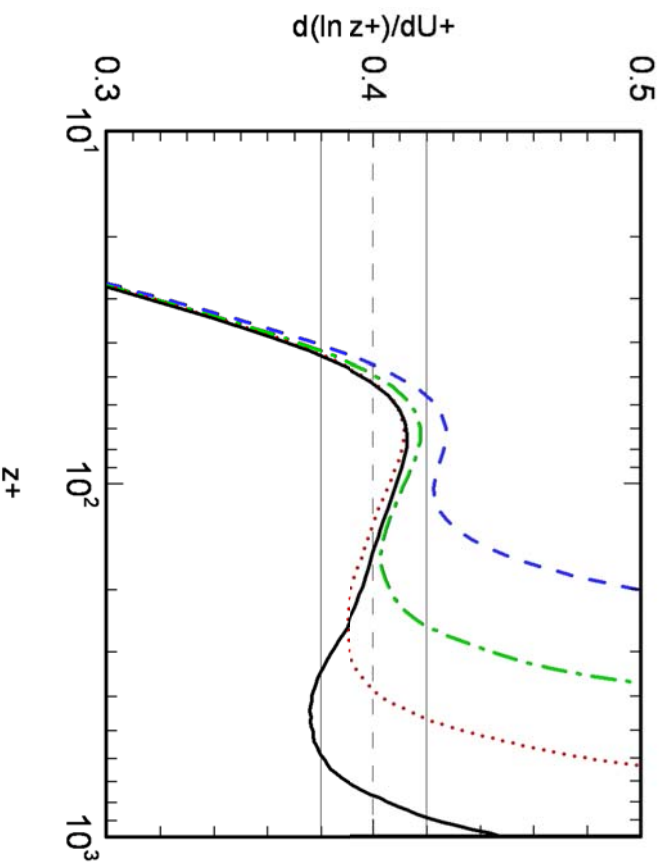


Log Law in Ekman-Layer DNS?

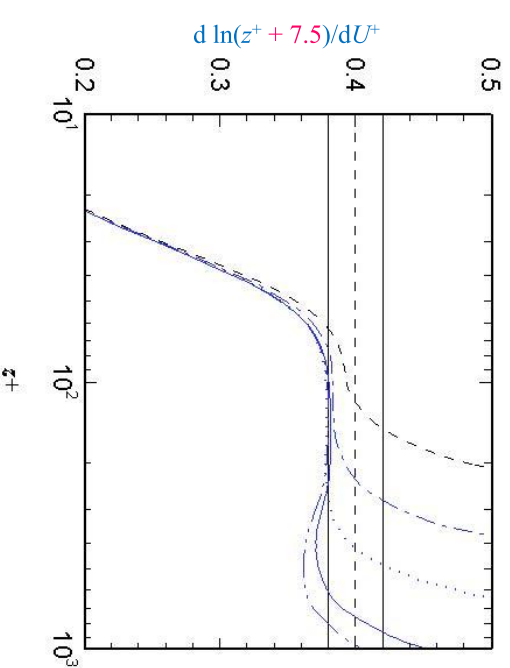


Ekman Reynolds numbers from 1000 to 2828. δ^+ scales like $Re^{1/6}$

Karman Measure in Ekman-Layer DNS



Karman Measure in Ekman-Layer DNS with Shift

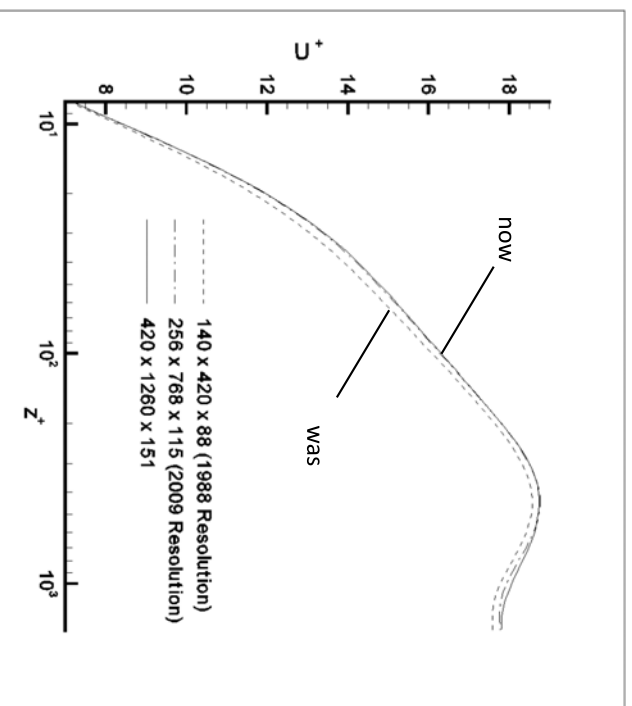


- Shifting to $\ln(z^+ + 7.5)$ creates a plateau at 0.38! (Main result of 2008 Phys Fluids paper.)

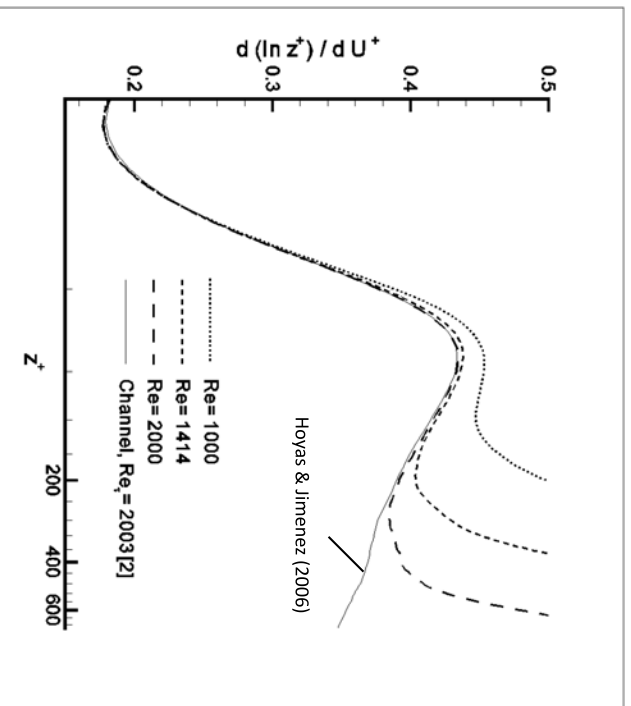
BUT, it was all too good to be true...

- Subsequent work has revealed 2008 Ekman DNS used Rules of Thumb for resolution (from Spalart's 1988 TBL) *that are not adequate*:
 - 1988 Resolution:
 - $\Delta x^+ = 20$, $\Delta y^+ = 7$, $n=10$ at $z^+ = 10$
 - 2009 Resolution:
 - $\Delta x^+ = 12$, $\Delta y^+ = 4$, $n=10$ at $z^+ = 6$

Resolution check: Ekman DNS at $Re=1000$



Karman measure for new Ekman DNS



New Ekman DNS:

- ☺ Agrees with channel, implies unique Law of the Wall $U^+=f(z^+)$.
- ☹ No sign of plateau in Karman Measure (and would require implausibly large offset to create one).
- ☹ Correction in Phys. Fluids **21**, 109901 (2009).
- ☹ Determination of κ via DNS now appears many years away.

Summary

- Couette-Poiseuille DNS implies the Log Law is very robust.
- Is it universal? *We'd like to think so, but time will tell...*
 - (An order of magnitude over Kim *et al* (1987) has been gained in Re , but κ is less certain than ever!)
 - NB. Recent "Variable κ " proposals = *death* of Log Law.
- Lessons from Ekman-layer DNS:
 - The general Law of the Wall (i.e. $U^+=f(z^+)$) is unique.
 - The Log Law is established only for $z^+ > 200$ at best (conventional ideas about Viscous, Buffer and beginning of Log layer need revision).
 - Using Standard Rules of Thumb regarding resolution requires caution, and does *not in-and-of-itself guarantee perfection at every level*.