LES of Turbulent Flows: Lecture 2 (ME EN 7960-003)

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Turbulent Flow Properties Review from Previous Properties of Turbulent Flows: Unsteadiness: 1. u=f(**x**,t) time 2. <u>3D:</u> contains random-like variability in space x_i (all 3 directions) 3. High vorticity: Vortex stretching — mechanism to increase the intensity of turbulence (we can measure the intensity of turbulence with the turbulence intensity => $\frac{\sigma_u}{\langle u \rangle}$) Vorticity: $\omega = \vec{\nabla} \times \vec{u}$ or $\omega_k = \epsilon_{ijk} \frac{\partial}{\partial x_i} u_j \hat{e}_k$ THE UNIVERSITY OF UTAH

Turbulent Flow Properties (cont.)

Properties of Turbulent Flows:

4. Mixing effect:

Turbulence mixes quantities with the result that gradients are reduced (e.g. pollutants, chemicals, velocity components, etc.). This lowers the concentration of harmful scalars but increases drag.

5. <u>A continuous spectrum (range) of scales:</u>





Kolmogorov's Similarity hypothesis (1941)

Kolmogorov's 1st Hypothesis:

- smallest scales receive energy at a rate proportional to the dissipation of energy rate.
- motion of the very smallest scales in a flow depend only on:

a) rate of energy transfer from small scales: $\epsilon \left| \frac{L^2}{T^3} \right|$

b) kinematic viscosity: $\nu \left[\frac{L^2}{T}\right]$

With this he defined the Kolmogorov scales (dissipation scales):

- length scale: $\eta = \left(\frac{\nu^3}{\epsilon}\right)^{\frac{1}{4}}$
- time scale: $au = \left(\frac{\nu}{\epsilon}\right)^{\frac{1}{2}}$
- velocity scale: $v = (\nu \epsilon)^{\frac{1}{4}}$

Re based on the Kolmolgorov scales => Re=1

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Kolmogorov's Similarity hypothesis (1941)

From our scales we can also form the ratios of the largest to smallest scales in the flow (using ℓ_o , U_o , t_o). Note: dissipation at large scales => $\epsilon \sim \frac{U_o^3}{\ell_o}$

• length scale:

$$\eta = \left(\frac{\nu^3}{\epsilon}\right)^{\frac{1}{4}} \sim \left(\frac{\nu^3 \ell_o}{U_o^3}\right)^{\frac{1}{4}} \Rightarrow \frac{\eta}{\ell_o^{1/4}} \sim \frac{\nu^{3/4}}{U_o^{3/4}} \Rightarrow \frac{\eta}{\ell_o} \sim \frac{\nu^{3/4}}{U_o^{3/4} \ell_o^{3/4}} \sim Re^{-3/4}$$

• velocity scale:

$$v = (\nu\epsilon)^{\frac{1}{4}} \sim \left(\frac{\nu U_o^3}{\ell_o}\right)^{\frac{1}{4}} \Rightarrow \frac{v}{U_o^{3/4}} \sim \frac{\nu^{1/4}}{\ell_o^{1/4}} \Rightarrow \frac{v}{U_o} \sim Re^{-1/4}$$

• time scale:

 $\tau = \frac{\eta}{v} \Rightarrow \frac{\tau}{t_o} \sim Re^{-1/2}$

For very high-Re flows (e.g., Atmosphere) we have a range of scales that is small compared to ℓ_o but large compared to η . As *Re* goes up, η / ℓ_o goes down and we have a larger separation between large and small scales.

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Kolmogorov's Similarity hypothesis (1941)

Kolmolgorov's 2nd Hypothesis:

In Turbulent flow, a range of scales exists at very high Re where statistics of motion in a range ℓ (for $\ell_o >> \ell >> \eta$) have a universal form that is determined only by ϵ (dissipation) and independent of ν (kinematic viscosity).

• Kolmogorov formed his hypothesis and examined it by looking at the pdf of velocity increments Δu .



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• What are structure functions ???

Important single point stats for joint variables

• <u>covariance:</u>

$$\operatorname{cov}(U_1, U_2) = \langle u_1 u_2 \rangle = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (V_1 - \langle U_1 \rangle) (V_2 - \langle U_2 \rangle) f_{12}(V_1, V_2) dV_2 dV_1$$

• Or for discrete data

$$\operatorname{cov}(U_{1}, U_{2}) = \langle u_{1}u_{2} \rangle = \frac{1}{N-1} \sum_{j=1}^{N} (V_{1j} - \langle U_{1} \rangle) (V_{2j} - \langle U_{2} \rangle)$$

We can also define the correlation coefficient (non dimensional)

$$\rho_{12} = \frac{\langle u_1 u_2 \rangle}{\left[\langle u_1^2 \rangle \langle u_2^2 \rangle \right]^{\frac{1}{2}}}$$

- Note that $-1 \le \rho_{12} \le 1$ and negative value mean the variables are anticorrelated with positive values indicating a correlation
- **Practically speaking**, we find the PDF of a time (or space) series by:
 - 1. Create a histogram of the series (group values into bins)
 - 2. Normalize the bin weights by the total # of points

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Two-point statistical measures

- <u>autocovariance</u>: measures how a variable changes (or the correlation) with different lags $R(s) \equiv \langle u(t)u(t+s) \rangle$
- or the autocorrelation function

$$\rho(s) \equiv \left\langle u(t)u(t+s)\right\rangle / \left\langle u(t)^{2}\right\rangle$$

- These are very similar to the covariance and correlation coefficient
- The difference is that we are now looking at the linear correlation of a signal with itself but at two different times (or spatial points), i.e. we lag the series.
- Discrete form of autocorrelation:

$$\rho(s_{j}) = \frac{\sum_{k=0}^{N-1} (u_{k}u_{k+j})}{\sum_{k=0}^{N-1} (u_{k}^{2})}$$

 $N \neq 1$

We could also look at the cross correlations in the same manner (between two different variables with a lag).

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• Note that: $\rho(0) = 1$ and $|\rho(s)| \le 1$

Two-point statistical measures

- In turbulent flows, we expect the correlation to diminish with increasing time (or distance) between points:
- We can use this to define an ¹ Integral time scale (or space). It is defined as the time lag where the integral $\int \rho(s) ds$ converges. and can be used to define the largest scales of motion (statistically).



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Another important 2 point statistic is the structure function:

$$D_n(r) \equiv \left\langle \left[U_1(x+r,t) - U_1(x,t) \right]^n \right\rangle$$

This gives us the average difference between two points separated by a distance r raised to a power n. In some sense it is a <u>measure of the moments of the velocity</u> <u>increment PDF</u>. Note the difference between this and the **autocorrelation which is statistical linear correlation** (ie multiplication) of the two points.

Fourier Transforms

Alternatively, we can also look at turbulence in wave (frequency) space:

Fourier Transforms are a common tool in fluid dynamics (see Pope, Appendix D-G, Stull handouts online)

Some uses:

- Analysis of turbulent flow
- Numerical simulations of N-S equations
- Analysis of numerical schemes (modified wavenumbers)
- consider a periodic function f(x) (could also be f(t)) on a domain of length 2π
- The Fourier representation of this function (or a general signal) is:

$$f(x) = \sum_{k=-\infty}^{k=\infty} \hat{f}_k e^{ikx} \qquad \bigstar$$

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- where *k* is the wavenumber (frequency if *f*(*t*))
- f_k are the Fourier coefficients which in general are complex



Fourier Transform Applications

Energy Spectrum: (power spectrum, energy spectral density)

• If we look at specific *k* values from we can define:

$$E(k) = N \left| \hat{f}_k \right|^2$$

The square of the Fourier coefficients is the contribution to the variance by fluctuations of scale k (wavenumber or equivalently frequency)

where E(k) is the energy spectral density

- Typically (when written as) E(k) we mean the contribution to the turbulent kinetic energy (tke) = $\frac{1}{2}(u^2+v^2+w^2)$ and we would say that E(k) is the contribution to tke for motions of the scale (or size) k. For a single velocity component in one direction we would write $E_{11}(k_1)$.
- See supplement for more on Fourier Transforms



Kolmogorov's Similarity Hypothesis (1941)

- Another way to look at this (equivalent to structure functions) is to examine what it means for E(k) where E(k)dk = t.k.e. contained between k and k + dk
- What are the implications of Kolmolgorov's hypothesis for E(k)? $K41 \Rightarrow E(k) = f(k, \epsilon)$

By dimensional analysis we can find that: $E(k) = c_k \epsilon^{2/3} k^{-5/3}$

• This expression is valid for the range of length scales ℓ where $\ell_o >> \ell >> \eta$ and is usually called the inertial subrange of turbulence.

