

- stochastic approach to turbulence -



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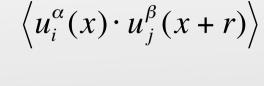
Joachim Peinke





turbulence

open question: to understand the correlations of the disorder of the turbulent field



for r = > 0 Reynolds stress

alternatively increments for spatial correlations

$$\vec{u}_r(x) = \vec{u}(x+r) - \vec{u}(x)$$

with u_r longitudinal and v_r transversal increments





statistics of turbulence

challenge to know - general n-scale statistics



$$p(\vec{u}_1, r_1; \vec{u}_2, r_2; ...; \vec{u}_n, r_n)$$

$$\left\langle \vec{u}_1^{\alpha_1} \cdot \vec{u}_2^{\alpha_2} \dots \vec{u}_n^{\alpha_n} \right\rangle$$

Known is

Kolmogorov
$$\langle u_r^3 \rangle = -\frac{4}{5} \varepsilon_r r + 6 v \frac{\partial}{\partial r} \langle u_r^2 \rangle$$

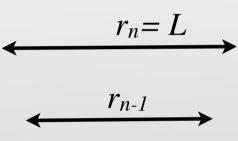
 $\langle u_r(x)^n \rangle \propto C_n r^{\xi_n}$

Karman
$$-r\frac{\partial}{\partial r}\left\langle u_{r}^{2}\right\rangle = 2\left\langle u_{r}^{2}\right\rangle - 2\left\langle v_{r}^{2}\right\rangle$$

statistics of turbulence

n-scale statistics

$$p(\vec{u}_1, r_1; \vec{u}_2, r_2; ...; \vec{u}_n, r_n)$$



$$\leftarrow r_{n-1} \rightarrow$$

$$\leftarrow r_{n-2} \rightarrow$$

$$\leftarrow$$

what are possible simplifications?

all increments at the same location

$$u_{r_i} =: u_i = u(x + r_i) - u(x)$$



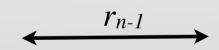
statistics of turbulence -2-

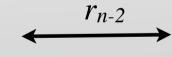
n-scale statistics

$$p(\vec{u}_1, r_1; \vec{u}_2, r_2; ...; \vec{u}_n, r_n)$$

what are possible simplifications?







$$r_{n-3}$$

formula of Bayes

$$p(\vec{u}_1, r_1; ...; \vec{u}_n, r_n) =$$

$$p(\vec{u}_1, r_1 | \vec{u}_2, r_2; ...; \vec{u}_n, r_n) p(\vec{u}_2, r_2; ...; \vec{u}_n, r_n)$$

simplification if:

$$p(\vec{u}_1, r_1 | \vec{u}_2, r_2; ...; \vec{u}_n, r_n) = p(\vec{u}_1, r_1 | \vec{u}_2, r_2)$$

or

$$p(\vec{u}_1, r_1 | \vec{u}_2, r_2; ...; \vec{u}_n, r_n) = p(\vec{u}_1, r_1)$$



statistics of turbulence -3-

simplification

(1)
$$p(\vec{u}_1, r_1 | \vec{u}_2, r_2; ...; \vec{u}_n, r_n) = p(\vec{u}_1, r_1 | \vec{u}_2, r_2)$$

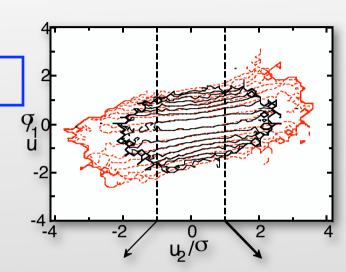
(2)
$$p(\vec{u}_1, r_1 | \vec{u}_2, r_2; ...; \vec{u}_n, r_n) = p(\vec{u}_1, r_1)$$

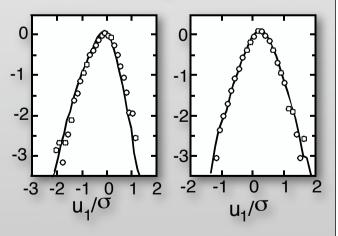
experimental test

experimental result:

$$p(u_1|u_2,u_3) = p(u_1|u_2)$$

- (I) holds
- (2) not







statistics of turbulence -4-

general n-scale statistics can be expressed by

$$p(\vec{u}_1, r_1; ...; \vec{u}_n, r_n) = p(\vec{u}_1, r_1 | \vec{u}_2, r_2) p(\vec{u}_2, r_2 | \vec{u}_3, r_3) ... p(\vec{u}_{n-1} | \vec{u}_n) p(\vec{u}_n, r_n)$$

and not

$$p(\vec{u}_1, r_1; ...; \vec{u}_n, r_n) \neq p(\vec{u}_1, r_1) p(\vec{u}_2, r_2) ... p(\vec{u}_n, r_n)$$

with cascades picture

$$\leftarrow r_n = L$$

Cascade a Markov process

$$r_{n-1}$$
 r_{n-2}
 r_{n-3}



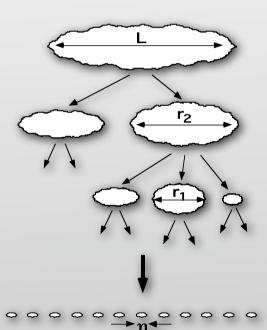


stochastic cascade process

idea of a turbulent cascade:

large vortices are generating small ones





 $\partial_r u_r$ $\partial_r p_r(u_r)$

=> stochastic cascade process evolving in r





stochastic cascade process - 2 -

summary: characterization of the disorder by joint n-scale statistics by a stochastic process,

I. proof of Markov properties

$$p(\vec{u}_1, r_1 | \vec{u}_2, r_2; ...; \vec{u}_n, r_n) = p(\vec{u}_1, r_1 | \vec{u}_2, r_2)$$

2. estimation of the Kramers Moyal coefficients results in simplification:

$$D^{(n)}(u,r) = \lim_{\Delta r \to 0} \frac{1}{n! \cdot \Delta r} \int (\tilde{u} - u)^n p(\tilde{u}, r - \Delta r \mid u, r) d\tilde{u}$$

3. obtain information for the n-scale statistics by process equation (Fokker-Planck or Kolomogorov equation)

$$-\frac{\partial}{\partial r}p(u,r|u_0,r_0) = \left[-\frac{\partial}{\partial u}D^{(1)}(u,r) + \frac{\partial^2}{\partial u^2}D^{(2)}(u,r)\right] \cdot p(u,r|u_0,r_0)$$



stochastic cascade process -3-

I. property of a Markov process:

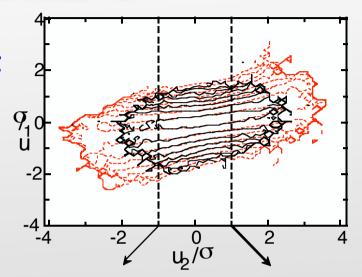
- evidence by conditional

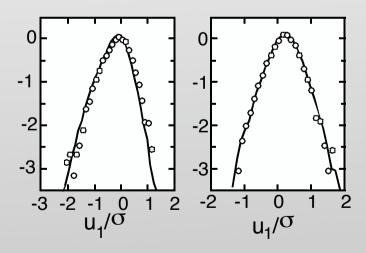
probability densities

$$p(u_1|u_2, ..., u_N) = p(u_1|u_2)$$

- experimental result:

$$p(u_1|u_2,u_3) = p(u_1|u_2)$$

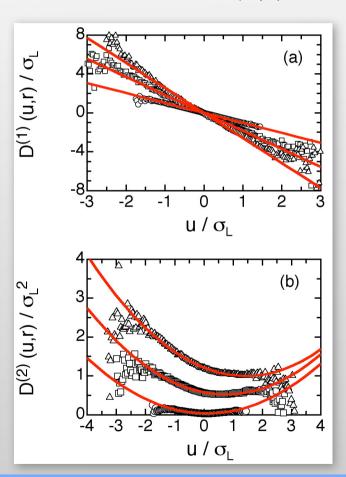






stochastic cascade process -4-

2. measured: $D^{(1)}(u,r)$ and $D^{(2)}(u,r)$



$$D^{(1)}(u,r) \cong \mathbf{v}(r) \ u(r)$$

$$D^{(2)}(u,r) \cong \alpha(r) + \delta(r) u(r) + \beta(r) u^2(r)$$

with the definition of (after Kol. 1931)

$$D^{(k)}(u,r) = \lim_{\Delta r \to 0} \frac{r}{k!\Delta r} M^{(k)}(u,r,\Delta r),$$

$$M^{(k)}(u,r,\Delta r) = \int_{-\infty}^{+\infty} (\tilde{u} - u)^k p(\tilde{u}, r - \Delta r | u, r) d\tilde{u}$$

stochastic cascade process -5-

measured Fokker-Planck equation

$$-\frac{\partial}{\partial r} \left\langle u_r^n \right\rangle = n \cdot \left\langle u_r^{n-1} D^{(1)}(u_r, r) \right\rangle + n \cdot (n-1) \left\langle u_r^{n-2} D^{(2)}(u_r, r) \right\rangle$$

- closed equation for structure functions if

$$\begin{split} D^{(1)}(u,r) &\cong \textcolor{red}{\gamma(r)} \ u(r) \\ D^{(2)}(u,r) &\cong \textcolor{red}{\alpha(r)} + \delta(r) \ u(r) + \textcolor{red}{\beta(r)} \ u^2(r) \end{split}$$

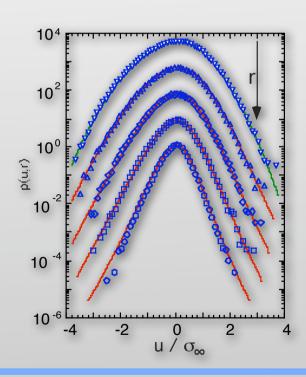


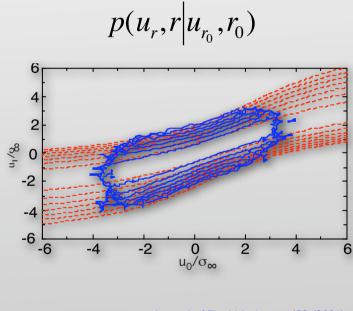


stochastic cascade process -6-

3. Verification of the measured Fokker-Planck equation

- numerical solution compared with experimental results
- => n-scale statistics

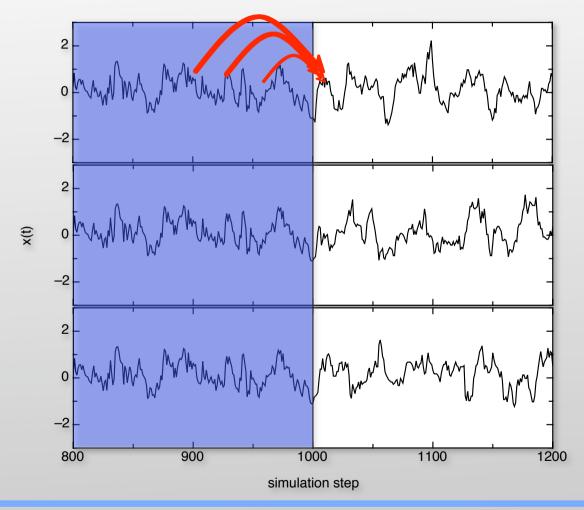




Journal of Fluid Mechanics 433 (2001) Phys. Rev E 76, 056102 (2007)



reconstruction of time series



use of increments alined to the right

$$p(u(x_{new}) \mid u(x_1) \dots, u(x_n))$$
 is given by

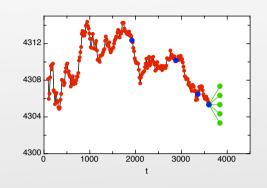
$$p(u_1, r_1; u_2, r_2; ...; u_n, r_{n-1})$$

Nawroth et al Phys. Lett. (2006)

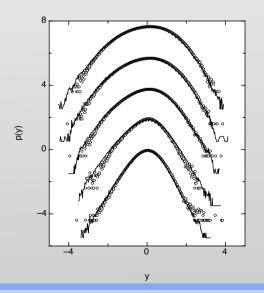


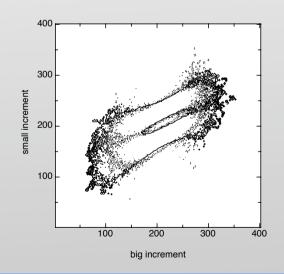
stochastic of reconstructed time series

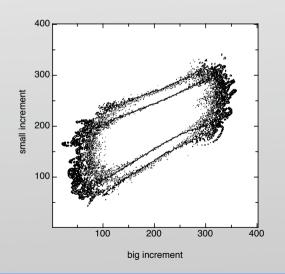
comparison of the statistics of the original and the modeled time series



same in the sense that $p(Q_0, \tau_0; Q_1, \tau_1; ...; Q_n, \tau_n)$ is the same





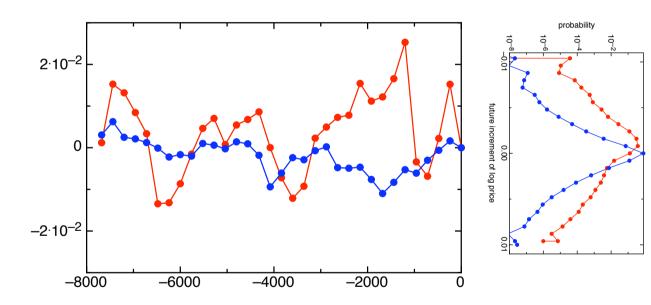






clustering of volatility for financial data

the statistics for the next step changes, for ,,quite" phase the variance is less than for ,,turbulent" phases (similar for turbulence)

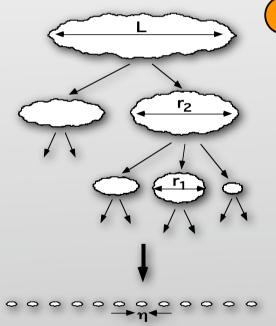




turbulent length scales

turbulent cascade: larger Re larger cascade range





integral length scale

$$Re = \left(\frac{L}{\eta}\right)^{3/4}$$



Taylor length scale



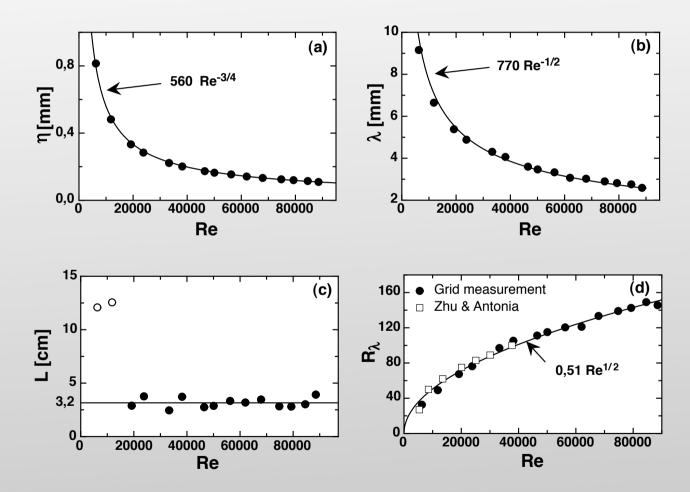
dissipation length scale





turbulent length scales

from grid experiments



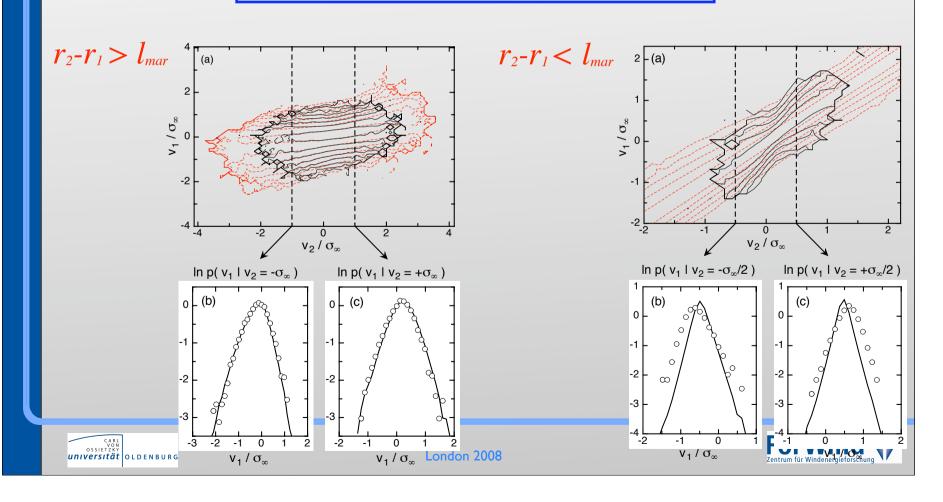




Einstein-Markov length

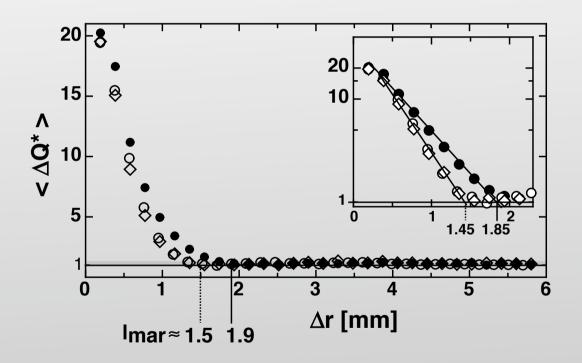
Einstein-Markov-length - a coherence length *l*_{mar}

$$p(\vec{u}_1, r_1 | \vec{u}_2, r_2; ...; \vec{u}_n, r_n) = p(\vec{u}_1, r_1 | \vec{u}_2, r_2)$$



Einstein-Markov length -2-

stochastic Wilcoxon test defines *l*_{mar}

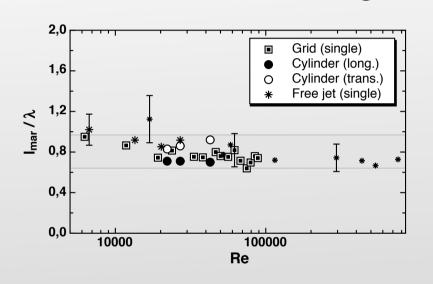


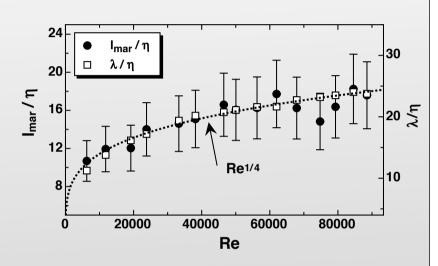




Einstein-Markov length -3-

Einstein-Markov length *lmar* a new coherence length





- is about the Taylor length
- is like the maximal dissipation length proposed by Yakhot
- dissipation causes memory
- degree of freedom L/l_{mar} like $Re^{1/2}$





Einstein - Markov Length

A. Einstein Ann. Phys. 17, 549 (1905)

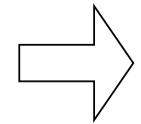
5. Über die von der molekularkinetischen Theorie der Wärme geforderte Bewegung von in ruhenden Flüssigkeiten suspendierten Teilchen; von A. Einstein.

§ 4. Über die ungeordnete Bewegung von in einer Flüssigkeit suspendierten Teilchen und deren Beziehung zur Diffusion.

Wir gehen nun dazu über, die ungeordneten Bewegungen genauer zu untersuchen, welche, von der Molekularbewegung der Wärme hervorgerufen, Anlaß zu der im letzten Paragraphen untersuchten Diffusion geben.

Es muß offenbar angenommen werden, daß jedes einzelne Teilchen eine Bewegung ausführe, welche unabhängig ist von der Bewegung aller anderen Teilchen; es werden auch die Bewegungen eines und desselben Teilchens in verschiedenen Zeitintervallen als voneinander unabhängige Vorgänge aufzufassen sein, solange wir diese Zeitintervalle nicht zu klein gewählt denken.

Wir führen ein Zeitintervall τ in die Betrachtung ein, welches sehr klein sei gegen die beobachtbaren Zeitintervalle, aber doch so groß, daß die in zwei aufeinanderfolgenden Zeitintervallen τ von einem Teilchen ausgeführten Bewegungen als voneinander unabhängige Ereignisse aufzufassen sind.





complexity of turbulence

thermodynamical (nonequilibrium) interpretation

- the Fokker- Planck or Kolmogov equation gives access

ideal gas

state vector

$$ec{q} = \left(egin{array}{c} ec{x} \ ec{p} \end{array}
ight)$$

n- particle description

$$p(q_1, q_2, ..., q_n)$$

single particle approximation

$$p(q_1,...,q_n)=p(q_1)*...*p(q_n)$$

Boltzmann equation

$$\partial_t p(q_i) = \dots$$

isotropic turbulence

state vector u_{l} ;

n- scale statistics

$$p(u_r_0, u_{r_1}, ..., u_{r_n})$$

Markov property

$$p(u_{r0},.., u_{rn}) = p(u_{r0}|u_{r1})*....$$

* $p(u_{rn-1}|u_{rn}) p(u_{rn})$

Fokker-Planck equation

$$-r\partial_r p(u_r|u_{r0}) = L_{FP} \ p(u_r|u_{r0})$$





turbulence: new insights

Einstein- Markov-length - a coherence length statistics of longitudinal and transversal increments universality of turbulence:

role of transfered energy e_r :

fusion rules $r_i = r_{i+1}$ (Davoudi, Tabar 2000; L'vov, Procaccia 1996)

passive scalar (Tutkun, Mydlarski 2004)





END





Langevin equations from time series

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after Pope and Ching

S. B. Pope and E. S. C. Ching, Phys. Fluids A 5, 1529 (1993).

$$p(x) = \frac{N'}{\langle\langle \dot{x}^2 | x \rangle\rangle} \exp \left[\int_x \frac{\langle\langle \ddot{x} | u \rangle\rangle}{\langle\langle \dot{x}^2 | u \rangle\rangle} du \right],$$

Stat. Solution of Fokker Planck

$$\frac{\partial p(x,t)}{\partial t} = -\frac{\partial}{\partial x} [A(x)p(x,t)] + \frac{1}{2} \frac{\partial^2}{\partial x^2} [B(x)p(x,t)],$$

$$p(x) = \frac{N}{B(x)} \exp\left[2\int_{x} \frac{A(u)}{B(u)} du\right],$$

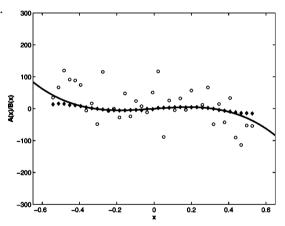


FIG. 3. Comparison among the estimated values of the ratio $\langle\langle\ddot{x}^2|x\rangle\rangle/\langle\langle\dot{x}^2|x\rangle\rangle$ using Sokolov's formulas (open circles), the ratio of the estimated drift and diffusion terms, $\langle\Delta x\rangle/\langle\Delta x^2\rangle$, using Eqs. (4) and (5) (solid diamonds), and the theoretical value, A(x)/B(x) (solid line), for the pitchfork bifurcation process.



see also http://www.physik.uni-oldenburg.de/hydro/20660.html

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