Conditional Averages and Probability Density Functions in the Passive Scalar Field

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Two conditional averages for the increment $\Delta(x,x+r) \equiv T(x)-T(x+r)$ of the scalar field are estimated for DNS data: $H(\Delta) = \langle \nabla^2 \Delta(x,x+r) | \Delta(x,x+r) \rangle$ and $G(\Delta) = \langle |\nabla \Delta(x,x+r)|^2 | \Delta(x,x+r) \rangle$. The probability density function $P(\Delta)$ for $\Delta(x,x+r)$ is also calculated. If $\Delta, P(\Delta), H(\Delta)$, and $G(\Delta)$ are expressed as $\theta, P(\theta), h(\theta)$, and $g(\theta)$ in a dimensionless form, the data analysis has revealed interesting simple relationships between them: (1) $g(\theta)P(\theta) \sim \exp(-\int_0^\theta dah(u)/g(u))$ derived by Ching and Kraichnan was numerically confirmed perfectly. (2) $g(\theta)P(\theta)$ pass through an almost common point for any scale ranging from the dissipative to the inertial, to the energy containing scale, (3) by using fitting parameters $\Lambda$ and $\beta$ as a function of $\theta$, $h(\theta)/g(\theta)$ is satisfactorily fitted by $(\Lambda/\beta) \tanh(\theta \beta)$, and then $g(\theta)P(\theta) \sim (\cosh \theta \beta)^{-\Lambda/\beta^2}$. These relations may be helpful to construct a turbulence model.

KEYWORDS: conditional averages, pdf, passive scalar field, anomalous scaling

§1. Introduction

Since Kraichnan’s work, the anomalous scaling of the passive scalar field has become a prototype of a study of the intermittency problem of turbulence. The scaling is evaluated by the scaling exponents of the $n$th order structure functions $S_n(r)$ of the scalar field $T(x)$, which are expressed as

$$S_n(r) = \langle \Delta(x,x+r)^n \rangle,$$

(1.1)

where $\Delta(x,x+r)$ is the increment of the scalar field denoted by

$$\Delta(x,x') = T(x) - T(x').$$

(1.2)

If the scalar filed is advected by the $\delta$-correlated Gaussian velocity field, $S_n(r)$ was shown to be governed by the following simple equation in the steady state

$$D_n(r) = J_n(r),$$

(1.3)

where $D_n(r)$ is the inertial term expressed by

$$D_n(r) = -r^{1-d} \frac{\partial}{\partial r} \left( a(r)r^{d-1} \frac{\partial}{\partial r} S_n(r) \right),$$

(1.4)

where $a(r)$ is the eddy diffusivity, and $d$ the dimension of the system. $J_n(r)$ is the dissipative term defined by

$$J_n(r) = n\kappa \langle \Delta(x,x')^{n-1} \Lambda(x,x') \rangle,$$

(1.5)

where

$$\Lambda(x,x') = (\nabla^2_x + \nabla^2_x) \Delta(x,x').$$

(1.6)

In eq. (1.5) $\kappa$ is the molecular diffusion coefficient and $x'$ is put equal to $x+r$.

In order to close the equation (1.3), one has to make a certain approximation on $J_n(r)$. The ansatz, which was employed by Kraichnan, is that the conditional average of the field $\Lambda(x,x')$ with a value of $\Delta(x,x')$ being fixed is proportional to $\Delta(x,x')$ itself, i.e.

$$H(\Delta(x,x')) = \langle \Lambda(x,x') | \Delta(x,x') \rangle = f_1(r) \Delta(x,x'),$$

(1.7)

where $f_1(r) = J_2(r)/(2\kappa S_2(r))$. Under this ansatz $J_n(r) = nJ_2(r)S_n(r)/(2S_2(r))$, which enables us to close the equation (1.3). Although this ansatz was examined in simulation and experiment, the definite confirmation has not been reached yet. Hence, it is interesting to examine to what extent of $\Delta$ the linear relation (1.7) is satisfied.

Another conditional average is conceivable. Making a partial integration in $J_n(r)$ with respect to $x$, we have

$$J_n(r) = -n(n-1)n\kappa \langle \Delta(x,x+r)^{n-2} \Sigma(x,x+r) \rangle,$$

(1.8)

where

$$\Sigma(x,x+r) = |\nabla_x(T(x) - T(x+r))|^2.$$  

(1.9)

Now we introduce another conditional average of $\Sigma(x,x+r)$ with a fixed value of $\Delta(x,x+r)$ as

$$G(\Delta(x,x+r)) = \langle \Sigma(x,x+r) | \Delta(x,x+r) \rangle.$$  

(1.10)

Although the linear ansatz has not been proved yet, it might have an origin in the dynamics of $\Delta(x,x+r)$. On the other hand, the conditional average (1.11) might
be closely related to the conservation of scalar dispersion, since Σ(\(x, x + r\)) represents the scalar diffusion rate. Which average is more appropriate will be judged through the examination of the result. This is the first aim of the present paper.

The second motivation of examining those averages is the recent argument by Ching and Kraichnan [11] which enables us to combine \(H\) and \(G\) to yield the probability distribution function (pdf) for \(\Delta(x, x')\). They showed that the pdf is related to the \(H/G\) in the homogeneous situation. We will confirm the relationship perfectly later in this paper. In this sense the conditional averages \(H\) and \(G\) are expected to shed light on the pdf of the increment of the scalar field and the issue of intermittency of turbulence.

Although the problem of the anomalous scaling should be ideally investigated in the randomly generated velocity fields, the generation of such fields is very difficult. Thus, the above conditional averages are examined in the usual Navier-Stokes turbulence in the present paper.

Finally we emphasize that the present paper is not directly devoted to the study of the recent issue of the anomalous scaling. One of the important features in the study is the introduction of the conditional average of spatial derivatives of the scalar field with a given value of its difference. Combination of the conditional average with the equation of motion and/or some kinematical constraints yields various check points to examine the pdf for the scalar difference, which would lead us to a more precise knowledge of the pdf. The point of view is not limited only to the case for the scalar field advected by the velocity field white in time. In this sense, we examine the statistics of the scalar field advected by the Navier-Stokes turbulence from the different point of angle with the help of the methods developed in the area of the anomalous scaling.

The present paper is organized as follows. In the following section the algorithm of the simulation as well as the statistical property of the velocity field is given. In §3 the pdf’s of the fields \(\Delta\) and \(\Lambda\) are presented with emphasis on their scale and Reynolds number dependence. To what extent the linear hypothesis of \(H\) is valid is examined in §4 and it is shown that it does not hold for large values of \(\Delta\), which correspond to the structure functions of orders larger than three. The deviation from the linearity is also discussed. In §5 the dependence of \(G\) on \(\Delta\) is investigated; it is described by a simple fitting formula, which is valid up to structure functions of large orders. Section 6 is devoted to the pdf of \(\Delta\) derived from the combination of \(H\) and \(G\). The numerically obtained pdf agrees with the observed one perfectly. The remarkable findings about \(H/G\) and \(P(\Delta)G\) are given and the results are analytically expressed using parameters. In §7 the useful findings are summarized to help to construct a turbulence model.

§2. Simulation

The simulation was done with resolutions of \(256^3\) (Run 1 and 2) and \(512^3\) (Run 3) mesh points for several values of molecular viscosity using the NAL Numerical Wind Tunnel [13]. The velocity and scalar field are simulated at the same time in \(4\pi\) cyclic box by making use of the Navier-Stokes equation and the advection equation. The Prandtl number is set equal to unity. The turbulence decays from the same initial spectrum for velocity and scalar fields

\[
\frac{16}{3} \sqrt{\frac{2}{\pi}} k^4 \exp(-2k^2). \tag{2.1}
\]

The data were processed at \(t=10\), when the energy dissipation rate reaches the largest value through the simulation, but the scalar one has passed the largest value already. At that time the Taylor microscale Reynolds number \(R_\lambda\) ranges from 89.5 for Run 1 to 120 for Run 3. The inertial region is identified from the scalar spectrum and the requirement of the relation \(T_r^2 u_r \sim r\), where \(T_r\) and \(u_r\) are the scalar and velocity increment over \(r\). The related data are summarized in Table 1.

|   | \(R_\lambda\) | \(k_{\text{max}}\eta\) | \(\lambda/\eta\) | \(L_p/\eta\) | Inertial range         |
|---|--------------|----------------|----------------|-------------|------------------------|
| Run 1 | 89.5          | 1.27           | 18.6           | 63.4        | \(7 < r/\eta < 31\)    |
| Run 2 | 120          | 0.716          | 21.5           | 68.4        | \(12 < r/\eta < 46\)   |
| Run 3 | 120          | 1.42           | 21.8           | 103         | \(14 < r/\eta < 51\)   |

Since the resolution for Run 1 and 3 are sufficient, the data associated with Run 1 and 3 are mainly discussed. The inertial region of Run 1 is located in \(31 > r/\eta > 7\) for Run 1 and \(51 > r/\eta > 14\) for Run 3, \(\eta\) being the Kolmogorov dissipation scale. The dissipation spectrum is the largest at about \(r/\eta = 5\) for both runs, so that the scale \(r/\eta = 5\), which will appear often later, is in the dissipative region.

Since Run 2 and Run 3 have the same Reynolds number with different resolution, the comparison of the result of Run 2 with that of Run 3 is interesting to know the effect of the difference of the resolution.

§3. The pdf’s of the fields \(\Delta_r\) and \(\Lambda_r\)

In this section we will be concerned with the pdf’s for the scalar field increment \(\Delta_r(x) = \Delta(x, x + r)\) over two points separated by \(r\) and the Laplacian field increment \(\Lambda_r(x) = \Lambda(x, x + r)\). For simplicity we introduce the
normalized dimensionless quantities such as

\[ \theta_r(x) = \frac{1}{\sqrt{S_2(r)}} \Delta_r(x), \]  
(3.1)

\[ \phi_r(x) = \frac{2\kappa \sqrt{S_2(r)}}{J_2(r)} \Lambda_r(x). \]  
(3.2)

The pdf's \( P(\theta_r) \) and \( P(\phi_r) \) of the fields \( \theta_r \) and \( \phi_r \) are obtained for various values of \( r \) ranging from the dissipative scale to the upper inertial one. From now on, we suppress the subscript \( r \) without otherwise stated. The pdf's for Run 3 are depicted in Fig.1, although the same ones could be drawn for Run 1; \( r/\eta = 20 \) in the inertial region in Fig.1(a), and \( r/\eta = 5 \) in the dissipative region in Fig.1(b). The vertical quantities as well as the horizontal ones are normalized. The solid line denotes \( P(\theta) \), while the broken line does \( P(\phi) \). The dotted line stands for the Gaussian distribution. In order to emphasize which part of the fluctuations contributes to the \( n \)th order structure functions, the arrows with \( P_n^+ \) and \( P_n^- \) are inserted: \( P_n^+ \) signifies a position where \( P(\theta)^n \) is the largest for \( \theta > 0 \) and \( P_n^- \) does a position where \( P(\theta)^n \) is the largest for \( \theta < 0 \).

The scale-dependence of pdf's of \( \theta \) and \( \phi \) is different. The frequency of large amplitudes of \( \theta \) becomes less as \( r \) increases, which implies that the degree of the intermittency measured from \( P(\theta) \) is weaken with the increase of the scale \( r \). On the other hand, the pdf of \( \phi \) does not vary so much as compared with that of \( \theta \) even when the scale changes. This difference is consistent with the fact that \( \theta_r \) represents the fluctuations of scale \( r \), while \( \phi_r \) does the ones of dissipative scale.

In order to show the Reynolds number dependence of both pdf's, we depicted \( P(\theta) \) at \( r/\eta = 20 \) for Run 1 and Run 3 in Fig.2(a) and \( P(\phi) \) in Fig.2(b). The solid line is for \( R_3 = 89.5 \) of Run 1, and the broken line is for \( R_3 = 120 \) of Run 3, together with the dotted line for Gaussian. The large components of \( \theta \) and \( \phi \) become more frequent as \( R_\lambda \) increases.

When one compares \( P(\phi) \) to \( P(\theta) \) of the same size \( r \), for instance by inspection of Fig.1(a), one notices that \( \phi \) is more frequent than \( \theta \) for large amplitudes and less frequent for intermediate ones for any \( r \). As \( r \) increases, the discrepancy between \( \phi \) and \( \theta \) becomes large, although figures associated with the other scales are not shown here. We suspected that the difference between both pdf’s would be responsible for the deviation of \( H \) from the linearity ansatz, which will be discussed in the following section, but we could not have found a clear reasoning for it so far.

The Run 2 and 3 have the same Reynolds number \( R_\lambda = 120 \), but are simulated on the different number of mesh points; the former is on \( 256^3 \) and the latter, on \( 512^3 \). Thus, the small scale resolution is less for Run 2 than for Run 3. \( P(\theta) \)'s for both runs are almost the same in the inertial region, but \( P(\phi) \)'s are considerably different from each other even in the inertial region. The curve of \( H(\Delta) \) for Run 2 is, however, found to be quite the same in the inertial range as Run 3. Hence the linearity ansatz (1.7) may be more robust in the inertial region than the pdf’s.

\section{§4. Examination of the linear ansatz}

The conditional average \( H \) is expressed in the dimensionless form as

\[ h(\theta_r) = H(\Delta_r) \frac{2\kappa \sqrt{S_2(r)}}{J_2(r)}. \]  
(4.1)

Note that \( h(\theta) \) is positively defined for \( \theta > 0 \) because \( J_2(r) \) is negative. Under this definition the linear ansatz (1.7) becomes

\[ h(\theta) = \theta. \]  
(4.2)

We investigated \( h(\theta) \) for various values of \( r/\eta \) for every run. The observed results are summarized. The linear relation (4.2) is approximately valid, but not in detail. Figure 3 is the curve of \( h(\theta) \) against \( \theta \) for \( r/\eta = 20 \) (the solid line) and \( r/\eta = 30 \) (the broken line) for Run 3. Note that the same type of curve holds for any inertial range separation for other runs. The dotted line stands for \( h(\theta) = \theta \). In the inertial region \( h(\theta) \) is slightly larger than \( \theta \) for small values of \( \theta \), while for larger values of \( \theta \) \( h(\theta) \) is considerably smaller than the linear relation as seen in Fig.3. Within the present simulations we did not see the Reynolds number dependence of \( h(\theta) \), although
The statistical reliability is assured up to $P_n$ with $n \geq 3$. This suggests that the values of $S_n$ with $n \geq 3$ do not rely on the linear hypothesis in the present simulation.

The function $h(\theta)$ must satisfy the conditions

$$\int_0^\infty (h(\theta) - \theta)\theta P(\theta)d\theta = 0, \quad (4.3a)$$

and

$$\int_0^\infty (h(\theta) - 1)\theta P(\theta)d\theta = 0, \quad (4.3b)$$

which are derived as follows. The definition of $H$ yields

$$\int \Delta H(\Delta) P(\Delta)d\Delta = \langle \Delta \Lambda \rangle = -\langle \Sigma \rangle = \frac{J_2}{2\kappa},$$

whose dimensionless form reduces to $\int \theta h(\theta) P(\theta)d\theta = 1$. On the other hand, $\theta$ is normalized so that $\int \theta^2 P(\theta)d\theta = 1$. Combining these two equations, we are led to the condition $\langle 1.3a \rangle$ and $\langle 1.3b \rangle$.

The condition $\langle 1.3a \rangle$ is important in the following sense. Since $P$ is a functional of $h(\theta)$ and positive definite, the function $h(\theta) - \theta$ must change its sign at least once if there exists any deviation of $h(\theta)$ from the linear ansatz. Function $\theta P(\theta)$ decays rapidly at large $\theta$ so that small deviation of $h(\theta)$ from the ansatz at small $\theta$ implies large deviation of $h(\theta)$ at large $\theta$. Equation (4.3) also can be used as an examination of the linear ansatz in that smaller deviation of $h(\theta) - \theta$ from 0 at small $\theta$ where the sample size is usually enough leads to smaller deviation at large $\theta$ where the sample size is poor. This is consistent with observation in Fig. 3.

In the dissipative range with $r/\eta = 5$, $h(\theta)$ is given in the inset of Fig. 3. The function $h(\theta)$ is slightly smaller than $\theta$ for small values of $\theta$, and larger than one for larger values of $\theta$ in contrast to the inertial range separations. The crossing of $h(\theta)$ with the linear relation $\theta$ occurs around $P_3$ for $r/\eta = 5$. This tendency is observed also in Run 1.

In order to know how much $h(\theta)$ is deviated from $\theta$ we evaluate a variance

$$\sigma = \int_{-\infty}^{\infty} (h(\theta) - \theta)^2 P(\theta)d\theta \quad (4.4)$$

for Run 3. At $r/\eta = 5$ in the dissipative region $\sigma = 0.008$. As $r$ increases, it decreases and reaches zero at $r/\eta = 10$, where the linear ansatz is good. As $r$ increases further in the inertial range, $\sigma$ grows; $\sigma = 0.01$ at $r/\eta = 30$.

The observation that $h(\theta)$ deviates from the linear relation for $n \geq 3$ in the inertial region is understood in the following way. The ansatz (1.7) is the projection of the field $\Lambda(x, x')$ on the field $\Delta(x, x')$. The coefficient $f_1(r)$ is obtained by multiplying eq.(1.7) by $\Delta(x, x')$ and taking the average over the distribution of $\Delta(x, x')$:

$$f_1(r)S_2(r) = \langle \Delta(x, x')H(\Delta(x, x')) \rangle = \frac{J_2(r)}{2\kappa}, \quad (4.5)$$

from which we have $f_1(r) = J_2(r)/(2\kappa S_2(r))$. The calculation of $J_2(r)$ is the same as that of the projection
coefficient. Since $J_2(r)$ is directly related to $S_2(r)$, it is naturally expected that the linear ansatz including the definition of $f_1(r)$ is well satisfied at $n = 2$, i.e. around $P_2$. Let us go to larger values of $n$. In the calculation of $J_n(r)$ the field $\Lambda(x, x')$ must be projected on the space where the quantity $\Delta(x, x')^{n-1}$ is dominant. Such a space is considerably localized, so that the projection coefficient $f_1(r)$ is not suitable. $f_1(r)$ must be replaced by

$$\tilde{f}_1(r) = \frac{\tilde{J}_2(r)}{2n\tilde{S}_2(r)}, \quad (4.6)$$

where $\tilde{J}_2(r)$ and $\tilde{S}_2(r)$ are computed over such an enhanced space. Since the equation for the scalar field is linear, one may expect that the ratio of $\tilde{J}_2(r)$ to $\tilde{S}_2(r)$ is constant everywhere. This would be correct if the scalar field did not move in space and the cascade occurred locally in space. While the strong fluctuations of scale $r$ cascade down to the dissipation scale where the field $\Lambda$ is mainly contributed, they will diffuse to the less intense space through the convection term due to the velocity field and the viscous term due to the molecular viscosity.

Although $\tilde{J}_2(r)$ is larger than $J_2(r)$, a ratio of $\tilde{J}_2(r)$ to $J_2(r)$ is less than that of $\tilde{S}_2(r)$ to $S_2(r)$. Hence $f_1(r)$ is expected to be less than $f_1(r)$, implying that $h(\theta)$ is lower than $\theta$. On the other hand, weak fluctuations will be affected in the opposite way, which may be responsible for the positive deviation from the linearity.

The above interpretation can be supported more directly by investigating the conditional average of the following identity with a fixed value of $\Delta(x, x + r)$:

$$\Delta(x, x + r)\nabla^2 \Delta(x, x + r) = \nabla \cdot (\Delta(x, x + r)\nabla \Delta(x, x + r)) - |\nabla \Delta(x, x + r)|^2.$$

Using the notation introduced previously, we have

$$\langle \nabla \cdot (\Delta(x, x + r)\nabla \Delta(x, x + r)) | \Delta(x, x + r) \rangle = \Delta(x, x + r)H(\Delta(x, x + r)) + G(\Delta(x, x + r)), \quad (4.7)$$

where the left hand side is the conditional average of the spatial transfer rate with fixed $\Delta(x, x')$. Since the two conditional averages on the right hand side are numerically evaluated, the dimensionless conditional transfer rate $t(\theta)$ is calculated as given in Fig.4. The region with positive values of $t(\theta)$ is characterized by the inflow, while that with negative values is by the outflow. About $P_2$, $t(\theta) \sim 0$ as expected. For large values of $\Delta(x, x')$ the outflow is dominant. Hence $\tilde{f}_1(r) < f_1(r)$ is anticipated in such a region. For small values of $\Delta(x, x')$ the inflow is dominant, so that the opposite tendency is expected.

§5. Examination of the second conditional average

We studied $G(\Delta_r)$, the conditional average of $\Sigma(x, x + r)$ with $\Delta(x, x + r)$ fixed. For simplicity we express $G$ in the dimensionless form

$$g(\theta_r) = -\frac{2\kappa}{J_2(r)} G(\Delta_r). \quad (5.1)$$

The analytic expression $g(\theta)$, if possible, is very useful to discuss the pdf $P(\theta)$, since it is directly related to the pdf as shown in the following section. Hence we try to investigate the functional form of $g(\theta)$ in detail.

First we plot $g(\theta)$ against $t^2$. The reason for it is that there must be a certain conservation relation between the scalar dispersion $2\theta$ and the scalar diffusion rate $q(\theta)$ for large values of $\theta$. Figure 5 is such a plot for $|\theta| < 5$. According to it, $g(\theta)$ may be fitted by a quadratic form

$$g(\theta) = g_0(r) + g_1(r)\theta^2 \quad (5.2)$$

for large values of $\theta$. For $r/\eta = 5$ in the dissipative region the curve is convex, so that $(5.2)$ is not valid; $g(\theta) \sim \theta^\alpha$ with $\alpha < 2$. In the inertial region $r/\eta = 15$ and $20$ we can see the considerable range fitted by $(5.2)$; the dotted straight line drawn on the curve for $r/\eta = 15$ is expressed by $1.7 + 0.12\theta^2$. The linearity is remarkable in the range between $P_3$ and $P_5$ (which are calculated for $r/\eta = 15$), where the linear ansatz $(1.7)$ fails, as seen in Fig.3. Such a fitting range becomes short as $r$ increases.

By fitting the formula $(5.2)$ to the curve of $g(\theta)$ of Fig.5 in the range $7 \leq \theta^2 \leq 13$, we estimated $g_0(r)$ and $g_1(r)$ as a function of $r$; they are depicted in Fig.6; the solid line is for $g_0$ and the broken one is for $10g_1$.

Both are slightly decreasing with $r$. Note that $g_1(r)$ is one order of magnitude less than $g_0(r)$.

In the region with small values of $\theta g(\theta)$ deviates from a fitting formula $(5.2)$. To clarify the point, we depicted $g(\theta)$ against $\theta$ in Fig.7 in the range $|\theta| \leq 2$. This figure indicates that the curve is described by a quadratic form in the vicinity of $\theta \sim 0$ and then, by the linear form for larger values of $\theta$ around 1. The linear region may be a transition region between the two different quadratic forms.

Before concluding this section we want to add that the function of $g(\theta)$ must satisfy the relation

$$\int (g(\theta) - 1) P(\theta) d\theta = \int (g(\theta) - \theta^2) P(\theta) d\theta = 0, \quad (5.3)$$

which is derived in a similar way to $(1.3a)$. From the
Fig. 5. The plot of $g(\theta)$ for various $r$ for Run 3. It is plotted against $\theta^2$. The linearity is quite good for $r/\eta = 15$ as seen from the dotted straight line $1.7 + 0.12\theta^2$ drawn on the curve for $r/\eta = 15$.

Fig. 6. The plot of $g_0(r)$ and $g_1(r)$, which are estimated from Fig. 5 in the range $7 \leq \theta^2 \leq 13$, as a function of $r/\eta$ in the inertial region for Run 3. The solid line is $g_0(r)$, while the broken line is $10g_1(r)$.

Fig. 7. The plot of $g(\theta)$ for various $r$ for Run 3. It is plotted against $\theta$ in the range of small values of $\theta$.

§6. Derivation of the PDF of $\theta$ in terms of $h(\theta)$ and $g(\theta)$

Ching and Kraichnan argued that $P(\theta)$ in the homogeneous system is expressed in terms of $h(\theta)$ and $g(\theta)$ through

$$P(\theta)g(\theta) = C \exp \left(-\int_0^\theta du q(u)\right),$$

(6.1)

where $C$ is a normalizing constant and

$$q(u) = \frac{h(u)}{g(u)}.$$  

(6.2)

Since the derivation of the relation is straightforward and short, we repeat their derivation for convenience. The extension of (4.7) to any $n$ yields

$$\left\langle \nabla_x \cdot \left(\Delta(x, x + r)^{n-1} \nabla_x \Delta(x, x + r)\right)\Delta(x, x + r)\right\rangle = \Delta^{n-1}(x, x + r)H(\Delta(x, x + r)) + (n - 1)\Delta^{n-2}(x, x + r)G(\Delta(x, x + r))$$

(6.3)

If the system is homogeneous in space, the average on the left hand side will be replaced by the spatial average, so that the left hand side vanishes. Then in the dimensionless form this equation becomes

$$\int d\theta \theta^{n-1} h(\theta) P(\theta) = (n - 1) \int d\theta \theta^{n-2} g(\theta) P(\theta)$$

$$= - \int d\theta \theta^{n-1} \frac{d}{d\theta} g(\theta) P(\theta).$$

(6.4)

The above relation must hold for any $n$, so that

$$h(\theta) P(\theta) = -\frac{d}{d\theta} g(\theta) P(\theta),$$

(6.5)

whose solution is $[\int 3]$. Hence the relation is expected to be exact for a spatially homogeneous system. Note that the similar expression for the pdf was proposed already by Yakhot in Bénard convection.

First, we present the observed result of $P(\theta)g(\theta)$ for Run 1 in Fig. 8(a) and for Run 3 in Fig. 8(b) for various values of $r$ ranging from $r/\eta = 5$ in the dissipative region to $r/\eta = 100$ in the energy containing region. The Gaussian pdf is also added. (1) The product form $P(\theta)g(\theta)$ looks simpler than $P(\theta)$ itself in Fig. 1. The fitting formula for $P(\theta)g(\theta)$ will be given later. (2) All curves pass through an almost common point. According to Fig. 8(a) and (b) the point is read as $(\theta = 1.5, P(\theta)g(\theta) = 0.13)$. When we magnified the crossing region of the curves, however, we noticed that the crossing does not occur at a single point. All the curves for the inertial region ranging from $r/\eta = 15$ to 50 (Run 3) cross at $\theta = 1.31$. On the other hand, the curves for the energy containing region and the dissipation region meet those for the inertial region at slightly larger values of $\theta$, i.e. about $(1.4 \sim 1.5)$. 

definition of $G$, its dimensionless form satisfies

$$\int g(\theta)P(\theta) d\theta = 1.$$
Second, we pay attention to the function $q(\theta) = \frac{h(\theta)}{g(\theta)}$. Substituting the measured values of $h(\theta)$ and $g(\theta)$ into $q(\theta)$, we numerically obtained $q(\theta)$ as a function of $\theta$; the function $q(\theta)$ for various values of $r/\eta$ for Run 3 is given in Fig.9. It increases linearly from the origin. The slope decreases with increasing $r$, but the decrease is not appreciable in the inertial region. For the separation in the dissipative region, $q(\theta)$ saturates and then, decreases very slowly. As $r$ increases, it does not saturate enough. The curves lose the symmetry a little when $|\theta|$ goes beyond 4. When $r$ increases, the considerable fluctuations are observed in the region with large values of $\theta$. Since $q(\theta)$ looks simpler than $h(\theta)$ and $g(\theta)$ themselves, the fitting formula for $q(\theta)$ is more suitable, as developed later.

Since the simulated system is considered to be homogeneous in space, the relation (6.1) is expected to hold. Substituting the calculated $q(u)$ into the right hand side of (6.1) and carrying out the numerical integration over $u$, we evaluated the right hand side of (6.1). Then the result was compared with the observed $P(\theta)$ multiplied by $g(\theta)$. The comparison was shown for $r/\eta = 5$ and $r/\eta = 20$ of Run 3 in Fig.10(a) and 10(b). The agreement is perfect.

Before going to discuss the analytical form for $q(\theta)$ and $g(\theta)P(\theta)$, we would like to consider which part of fluctuations contributes to $q(\theta)$, and $g(\theta)P(\theta)$ consequently. Let us start with small values of $\theta$, where $h(\theta) = \theta$ and $g(\theta) = \text{const} = c_0$. Hence, $q(u) = \frac{u}{c_0}$, leading to the Gaussian distribution according to (6.1) and the curvature is inversely proportional to $c_0$. Since $c_0$ decreases as $r$ decreases, as seen from Fig.7, the curvature around $\theta = 0$ is a decreasing function of $r$, in agreement with Fig.8.

The deviation from Gaussian begins to occur as $\theta$ grows. As $\theta$ increases, $h(\theta)$ is still close to $\theta$, but $g(\theta)$ is not constant. Figure 7 suggests that $g(\theta)$ is approximated by $g(\theta) = c_0 + c_1 \theta$ in that region, where $c_0$ increases and $c_1$ decreases with increasing $r$. Then

$$g(\theta)P(\theta) = C \exp \left( -\frac{\theta}{c_1} + \frac{c_0}{c_1} \log \frac{\theta + c_0/c_1}{c_0/c_1} \right). \quad (6.6)$$

The existence of $c_1$ makes the pdf deviate from Gaussian.
If the first term is dominant, the pdf is of the exponential type: the exponent $1/c_1$ increases with $r$, in consistent with the observation of Fig.8.

To conclude this section, we would like to propose an analytic expression of $q(\theta)$. The analytical expression, if possible, is helpful for constructing a turbulence model. The combined function $q(\theta)$ is simpler than $h(\theta)$ and $g(\theta)$ themselves. It may be approximately parametrized as

$$q(\theta) = \frac{A}{\beta} \tanh(\beta \theta),$$

(6.7)

where $A$ and $\beta$ depend on $r$. The most optimal fitting values for them are depicted in Fig. 11; the fitting was done in the symmetric region $|\theta| \leq 4$. The crosses stand for $A$; $A$ and $\beta$ are decreasing with $r$, while $A/\beta$ is increasing. Although the best fit for the entire region of $\theta$ is given in that figure, the local fitting is also possible. (For instance, the separate regions $\theta \leq 1$ and $\theta \geq 1$ would be fitted more appropriately.)

Substituting (6.7) into (6.1) yields

$$P(\theta)g(\theta) = C(\cosh(\beta \theta))^{-A/\beta^2},$$

(6.8)

where $C = \beta/B(A/2\beta^2, 1/2)$, with $B$ being a beta function. For small $\theta$, $P(\theta)g(\theta) \sim e^{-A\theta^2/2}$; i.e. Gaussian. The curvature decreases as $A$ when $\theta$ increases, in agreement with Fig.8. For large $\theta$, it becomes $e^{-A\theta^2/\beta}$, i.e. the pdf is of the exponential type, whose coefficient is $A/\beta$, which is an increasing function of $r$, also consistent with Fig.8.

In order to show how the fitting works, we would like to give $g(\theta)P(\theta)$ for $r/\eta = 10$ as an example in Fig.12, where $A = 1.44$ and $\beta = 1.60$ for $r/\eta = 10$ are employed from Fig.11. The solid line stands for the observed value of $g(\theta)P(\theta)$ in Fig.8(b) and the broken line does for the right hand side of (6.8) obtained after integrating $q(u)$ over $u$. The dotted line is given by (6.8) with selected $A$ and $\beta$. The agreement is very satisfactory.

![Fig. 11. The fitting parameters $A$ and $\beta$ as a function of $r$ for Run 3; the crosses stand for $\beta$, while the pluses do for $A$.](image)

![Fig. 12. The comparison of (6.8) to the observed function of $g(\theta)P(\theta)$ at $r/\eta = 10$. The solid line stands for the observed value of $g(\theta)P(\theta)$ and the broken line does for the right hand side of (6.8) obtained after integrating $q(u)$ over $u$. The dotted line is given by (6.8) with $A = 1.44$ and $\beta = 1.60$.](image)

§7. Discussion

When the equation (1.3) is expressed through the pdf, it takes the form

$$-r^{1-d} \frac{\partial}{\partial r} \left( a(r)r^{d-1} \frac{\partial}{\partial r} P(\Delta, r) \right) = -\kappa \frac{\partial}{\partial \Delta} \left( P(\Delta, r) H(\Delta, r) \right) = -\kappa \frac{\partial^2}{\partial \Delta^2} \left( P(\Delta, r) G(\Delta, r) \right).$$

(7.1)

If the velocity field is not $\delta$-correlated in time, the inertial term will be more complicated. In any case, however, $P(\Delta)$, $H(\Delta)$ and $G(\Delta)$ must satisfy similar equations. In this sense the information on $H$ and $G$ extracted from a study of the simulation is useful to construct a model for turbulence. For this purpose we summarize the useful points.

1. The combined form $q(\theta)$ is a more appropriate function for modeling, rather than $h(\theta)$ and $g(\theta)$ themselves.

2. Another combined form $P(\theta)g(\theta)$ is more appropriate than the pdf $P(\theta)$ itself. $P(\theta)g(\theta)$ is the weighted conditional average of the scalar diffusioon rate $g(\theta)$.

3. The curves $P(\theta)g(\theta)$ with varying $r$ drop on an almost common point.

Let us consider the meaning of $g(\theta)$. Take a specific scale $r$. Even if $\theta_r$ is fixed, there are many structures, so that the scalar diffusioon rate $|\nabla x \theta(x, x + r)|^2$ vary from structure to structure. After taking the average over structures, we have the coniddionally averaged dissipation rate $g(\theta_r)$.

The $g_r(\theta_r)P_r(\theta_r)$ is the average dissipation rate at scale $r$ weighted by the frequency $P_r(\theta_r)$ of $\theta_r$. Here we insert the subscript $r$ for the functions $g$ and $P$ to show their scale dependence explicitly. The present simulation indicates that $g_r(\theta_r)(\equiv g_r(\theta_r)P_r(\theta_r))$ takes the same value 0.13 at $\theta_r = 1.5$ irrespective of $r$. For $\theta_r \leq 1.5$ $g_r(\theta_r)$ increases with increasing $r$, while it decreases for $\theta_r \geq 1.5$. This is surprising, we think. Why? To show it, we con-
Consider two specific scales \( r_1 \) and \( r_2 \) such that \( r_1 < r_2 \). Then we can have a situation that \( \tilde{g}_{r_1}(\theta) \leq \tilde{g}_{r_2}(\theta) \) for \( \theta \leq \theta_c \), where \( \theta_c \) is a certain value, while \( \tilde{g}_{r_1}(\theta) \geq \tilde{g}_{r_2}(\theta) \) for \( \theta \geq \theta_c \). However, the present DNS suggests that this inequality occurs at the same \( \theta_c = 1.3 \) for any inertial separation \( r \), ranging from the dissipative to energy containing scale.

Finally we speculate the pdf in turbulence for large Reynolds numbers based on the results, although the present DNS decaying turbulence has only a limited interval of the inertial range. In this paper we observed the region where \( g(\theta) \propto \theta^2 \) for large values of \( \theta \). If such a term is dominant, and \( h(\theta) \) is \( \theta \), the function \( q(\theta) \) is proportional to \( \theta^{-1} \), yielding the power law distribution as suggested by Sinai and Yakhot\(^{15}\) and Ching\(^{16}\). Figure 5 indicates that the dominance of the square term over the constant term in \( g(\theta) \) becomes more enhanced as \( r \) decreases, so that the pdf of the smallest scale is expected to become more power-law like. However, in this simulation the pdf for the smallest scale is still of exponential type. The reason for it is that \( h(\theta) \sim \theta \), but \( g(\theta) \) is weaker than \( \theta^2 \), as seen in Fig.5, so that \( q(\theta) \) at \( r/\eta = 5 \) decreases slightly for large \( \theta \) (not as \( 1/\theta \)), as seen in Fig.9. In order to know the behavior of the pdf for large amplitude, therefore, the information of \( q(\theta) \) is more indispensable than that of individual conditional average \( h(\theta) \) and \( g(\theta) \).

\[\text{(1)}\]

\[\text{(2)}\]

\[\text{(3)}\]

\[\text{(4)}\]

\[\text{(5)}\]

\[\text{(6)}\]

\[\text{(7)}\]

\[\text{(8)}\]

\[\text{(9)}\]

\[\text{(10)}\]

\[\text{(11)}\]

\[\text{(12)}\]

\[\text{(13)}\]

\[\text{(14)}\]

\[\text{(15)}\]

\[\text{(16)}\]