Statistics of incremental averages of passive scalar fluctuations

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Abstract

Whereas statistical moments of differences of turbulent quantities measured over a given separation (viz. structure functions) have been extensively studied, statistics of incremental sums (or equivalently averages, e.g. $\Sigma \theta \equiv [\theta(x+r) + \theta(x)]/2$) of the same quantities have only been the subject of recent research. The present work investigates incremental averages of a turbulent passive scalar (temperature), measured in nearly homogeneous, and isotropic (passive and active) grid-generated turbulence, for turbulent Reynolds numbers in the range $94 \leq R_\lambda (\equiv u_{\text{rms}} \lambda/\nu) \leq 582$. The scalar field is generated by the action of the turbulent velocity field against an imposed mean temperature gradient. Following the approach of Mouri and Hori [Phys. Fluids, vol. 22, pp. 1-7, 2010] for the velocity field, we examine statistics of incremental averages of the passive scalar field as a function of separation (viz. incremental average structure functions) for different Reynolds numbers, comparing them with both the results of Mouri and Hori (2010), as well as the corresponding incremental average structure functions for the velocity field for the flows studied herein. While the statistics of $\Sigma \theta$ are primarily large-scale quantities, and would therefore be expected to be flow-dependent, they exhibit certain similarities to the statistics of incremental averages of velocity ($\Sigma u_\alpha$), measured both in the flow under consideration, as well as the different classes of flows studied by Mouri and Hori (2010). Lastly, we derive a scale-dependent evolution equation for the incremental average of the scalar field fluctuations, $\Sigma \theta$. We discuss its relationship to Yaglom’s four-thirds law for differences in passive scalar fluctuations and compare the results with the experimental data.
I. INTRODUCTION

In the years since Kolmogorov [1] published his theory of turbulent flows, there has been extensive experimental, numerical and theoretical research on the small scales of turbulent velocity fields. Following the subsequent extension of Kolmogorov theory to passive scalar fields by Obukhov [2] and Corrsin [3], there has been significant research on the structure of turbulent passive scalar fields, although to a lesser extent than that on their hydrodynamic counterparts. Turbulent passive scalar fields, which at first glance, might appear to be more straightforward than turbulent hydrodynamic fields (given the simpler nature of the advection-diffusion equation, as compared to the Navier-Stokes equations) are, in fact, quite distinct, exhibiting many unique behaviors. These include but are not limited to (i) the development of an inertial-convective subrange in homogeneous isotropic turbulence (e.g. Jayesh et al. [4]) at Reynolds and Péclet numbers well below those that permit a substantial separation of scales (the latter being the fundamental underpinning of Kolmogorov theory), (ii) strong, local anisotropy of small-scale quantities in flows that are anisotropic at large scales e.g., Tong and Warhaft [5], Mydlarski and Warhaft [6, 7], and (iii) larger levels of internal intermittency than observed in hydrodynamic fields [5, 6]. For further details on turbulent passive scalar fields, the reader is referred to the reviews of Sreenivasan [8], Warhaft [9], Shraiman and Siggia [10] and Dimotakis [11].

Given that Kolmogorov theory provides explicit predictions regarding their scaling, structure functions have played and important role and been the subject of numerous investigations of the small-scale statistics of turbulent fields. Small-scale statistics have been investigated by studying the structure functions, which are defined as the average of the difference between two values of a given quantity measured over a spatial (or temporal) interval, and raised to the power $n$. Using an arbitrary component of the turbulent velocity fluctuation ($u_\alpha$) as an example, an $n^{th}$-order structure function can be mathematically expressed as follows:

$$\langle (\Delta_r u_\alpha)^n \rangle \equiv \langle (u_\alpha(x_i + r_i, t) - u_\alpha(x_i, t))^n \rangle. \quad (1)$$

Structure functions provide information about the scaling of the small-scale turbulent fields. And although structure functions are sometimes claimed to be representative of the behavior of a given quantity at a scale $r$ ($\equiv |r| = (r_ir_i)^{1/2}$), where the Einstein summation
convention is only implied for Roman indices, not Greek ones, it is important to emphasize that structure functions also have contributions from scales smaller than $r$. To reinforce this point, note that the limit as $r$ tends to zero of the second-order structure function of $u_\alpha$ is zero, whereas its limit as $r$ tends to infinity is $2\langle u_\alpha^2 \rangle$, which is four times the turbulent kinetic energy ($\frac{1}{2}\langle u_\alpha^2 \rangle$) associated with the $u_\alpha$ component of velocity. Thus, a structure function contains contributions from all scales less than or equal to $r$, because the turbulent kinetic energy of a quantity has contributions from all scales. Recall that the turbulent kinetic energy can, after all, also be obtained from the integral of the power spectrum of a velocity fluctuation over all wavenumbers.

Given the extensive attention that has been paid to structure functions, a different, but complementary, statistic has been proposed to further our understanding of turbulent flows: two-point incremental averages of turbulent quantities, defined as follows:

$$\langle (\Sigma_i u_\alpha)^n \rangle \equiv \left\langle \left[ \frac{u_\alpha(x_i + r_i) + u_\alpha(x_i)}{2} \right]^n \right\rangle. \quad (2)$$

When referring to incremental averages over a separation $r$, we will use the phrase “incremental average structure functions” (IASFs), which are to be contrasted with traditional structure functions – like those defined by equation (1) – which we will refer to as “incremental difference structure functions” (IDSFs). The present work studies statistics of IASFs, with an emphasis on those of passive scalar fields. A brief comparison of the properties of IDSFs and IASFs (written for passive scalar fields) is given in Table I. Note that from here on in, the subscript “$r_i$” will be omitted (i.e. “$r$”), for the sake of concision.

Previous researchers have studied statistics of incremental averages, with Sreenivasan and Dhruva [12] being among the earliest to do so. Using atmospheric velocity measurements, they found that the probability density function of $\Sigma u$ to be nearly identical to the PDF of $u$. They also studied the expectations of incremental difference structure functions, conditioned on the velocity at the mid-point of an interval, which they noted gave the same results as conditioning on the incremental average velocity. Tatsumi and co-workers [13–15] theoretically studied the velocity sum PDF in homogeneous, isotropic turbulence using the cross-independence closure hypothesis. Statistics of incremental averages are also especially relevant to the (unresolved) question of whether large- and small-scale statistics of turbulent fields are statistically independent (e.g. Praskovsky et al. [16]). This question has been recently studied by Hosokawa [17], Kholmyansky and Tsinober [18], and Blum et al. [19].
Moreover, the energy transfer between two-point averages of a velocity field and its respective two-point differences was studied by Germano [20], who showed that the transfer can be viewed as being produced by the subgrid stress associated with the two-point average, or related to the classical, second-order IDSF. Finally, of particular interest to the present work is the work of Mouri and Hori [21], who extensively studied (longitudinal) incremental averages of the longitudinal and transverse velocity fluctuations ($\langle (\Sigma_r u)^n \rangle$ and $\langle (\Sigma_r v)^n \rangle$, respectively) in grid turbulence, a turbulent boundary layer, and a turbulent jet. Their work will serve as a reference for the results presented herein. In summary, given their relevance to both fundamental and applied issues in the study of turbulence, statistics of incremental averages merit further study.

<table>
<thead>
<tr>
<th></th>
<th>Incremental Difference Structure Functions</th>
<th>Incremental Average Structure Functions</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Definition:</strong></td>
<td>$\langle (\Delta \theta)^2 \rangle \equiv \langle (\theta(x_i + r_i, t) - \theta(x_i, t))^n \rangle$</td>
<td>$\langle (\Sigma \theta)^2 \rangle \equiv \langle [\frac{1}{2}(\theta(x_i + r_i) + \theta(x_i))]^n \rangle$</td>
</tr>
<tr>
<td><strong>Relationship to the autocorrelation, $\rho_{\theta \theta}(r)$:</strong></td>
<td>$\frac{\langle (\Delta \theta)^2 \rangle}{\langle \theta^2 \rangle} = 2(1 - \rho_{\theta \theta}(r))$</td>
<td>$\frac{\langle (\Sigma \theta)^2 \rangle}{\langle \theta^2 \rangle} = \frac{1}{2}(1 + \rho_{\theta \theta}(r))$</td>
</tr>
<tr>
<td><strong>Scaling:</strong></td>
<td>$\frac{\langle (\Delta \theta)^2 \rangle}{\langle \theta^2 \rangle} \propto Ar^{\gamma_{\theta}}$</td>
<td>$\frac{\langle (\Sigma \theta)^2 \rangle}{\langle \theta^2 \rangle} \propto 1 - Cr^{\gamma_{\theta}}$</td>
</tr>
</tbody>
</table>

TABLE I. Comparison of properties of the (traditional) incremental difference structure functions and those of incremental average structure functions, assuming statistical homogeneity in the $x$-direction.

To further contrast IDSFs and IASFs, we remark that a power-law scaling range (i.e., $\propto r^n$) is not expected to be (explicitly) observed in the statistics of incremental averages. However, in the limit of high Reynolds and Péclet numbers, the second-order incremental average structure function can be expected to scale as $1 - Cr^{\gamma_{\theta}}$, where $\gamma_{\theta}$ is the inertial-convective range scaling exponent of the second-order incremental difference structure function, $\langle (\Delta \theta)^2 \rangle$. This result stems from an easily-derivable relationship between the second-order IDSF and the second-order IASF [21]:

4
\[
\frac{\langle (\Sigma \theta)^2 \rangle}{\langle \theta^2 \rangle} = 1 - \frac{\langle (\Delta \theta)^2 \rangle}{4 \langle \theta^2 \rangle},
\] (3)

or, equivalently:

\[
\langle (\Sigma \theta)^2 \rangle + \frac{1}{4} \langle (\Delta \theta)^2 \rangle = \langle \theta^2 \rangle,
\] (4)

which can equivalently be derived from the relationships in Table 1 by eliminating \( \rho_{\theta \theta} \) and assuming statistical homogeneity in the \( x \)-direction — an assumption that will be relaxed in §3.

From equation (4), it can be concluded that motions at scales greater than or equal to \( r \) (as hypothesized by Mouri and Hori [21]) contribute to IASFs because (i) \( \langle \theta^2 \rangle \) has contributions from all scales \( (0 < r < \infty) \), and (ii) \( \langle (\Delta \theta)^2 \rangle \) is determined by motions at scales less than or equal to \( r \), as previously noted. This conclusion is directly related to the previous arguments pertaining to the scales on which IDSFs depend. Moreover, it explains the following result obtained by Hosokawa [17] for turbulence that is homogeneous in the \( x \)-direction:

\[
\langle (\Delta u)(\Sigma u)^2 \rangle = -\frac{1}{12} \langle (\Delta u)^3 \rangle,
\] (5)

which, in the inertial subrange, can be rewritten as:

\[
\langle (\Delta u)(\Sigma u)^2 \rangle = \frac{1}{15} \langle \epsilon \rangle r.
\] (6)

without an invocation of a correlation between large- and small-scale motions, as proposed by Hosokawa [17], and where \( \epsilon \) denotes the dissipation rate of turbulent kinetic energy, \( \frac{\nu}{2}(\partial u_i/\partial x_j + \partial u_j/\partial x_i)^2 \), where \( \nu \) is the fluid kinematic viscosity. Rather, the non-zero correlation between \( \Delta u \) and \( (\Sigma u)^2 \) arises from the fact that both have non-zero contributions at scale \( r \).

To conclude this section we state the main objectives of the present work, which are to (i) provide the first measurements (to our knowledge) of incremental averages of a turbulent scalar field, (ii) provide further insight on the structure of turbulent passive scalar fields by means of the aforementioned measurements, and (iii) theoretically and experimentally investigate the relationship between incremental averages of a turbulent scalar field and the velocity field that advects it. To these ends, we will study passive scalar fields generated by
a mean temperature gradient in grid turbulence (generated by both passive and active grids) over the Reynolds number range $94 \leq R_\lambda \leq 582$. In addition, we will present experimental measurements of incremental averages of the longitudinal and transverse velocity fields, and compare them with those of the scalar field. The remainder of this paper is organized as follows. The apparatus is described in §2. The results (of both the scalar and velocity fields) are presented and discussed in §3. Conclusions follow in §4.

II. APPARATUS

The measurements presented herein were made in the $0.91\text{m} \times 0.91\text{m} \times 9.1\text{m}$ low-speed, low-background-turbulence wind tunnel in the Sibley School of Mechanical and Aerospace Engineering at Cornell University. The experiments and apparatus have been described in Mydlarski and Warhaft [6] as well as Danaila and Mydlarski [22]. Therefore, the discussion herein of the apparatus will be relatively brief.

The flow is homogeneous, quasi-isotropic, grid-generated turbulence. One set of measurements employed a passive grid with a mesh length ($M$) of 10.16 cm (4 inches). Larger Reynolds (and Péclet) numbers were obtained by use of an active grid [23–25]. The mean temperature gradient is generated by a set of differentially heated ribbons placed in the wind tunnel plenum chamber [see 6, 26, 27]. The mean flow is in the $x$-direction while the mean scalar gradient is in the $y$-direction. Both the velocity and temperature fields are statistically homogeneous in the $z$-direction.

The longitudinal and transverse components of velocity were measured by a TSI 1241 X-wire probe with 3.05 µm diameter tungsten wires that were operated using Dantec 55M01 constant-temperature anemometers. The length-to-diameter ratio of each wire was approximately 200 and the inter-wire spacing was 0.5 mm. The turbulent temperature field was measured by cold-wire thermometry in conjunction with a TSI 1210 probe to which 0.63 µm diameter platinum core Wollaston wire was soldered. The temperature wire was located 0.5 mm from the X-wire and its length-to-diameter ratio varied from 500-650, depending on the flow. The hot-wire signals were calibrated using a modified King’s law with temperature-dependent coefficients [28] to account flow for the fluctuating temperature of the flow. Time series of the two components of velocity and temperature were high- and low-pass filtered, sampled at twice the Kolmogorov frequency, and digitized using a 12-bit analog-to-digital
converter. Spatial separations (used in the definition of the incremental average) are all in the longitudinal \((x)\) direction and were obtained using Taylor’s hypothesis. The associated flow parameters are outlined in Table II. For more details on the flows, the reader is referred to Mydlarski and Warhaft [6] and Danaila and Mydlarski [22].

<table>
<thead>
<tr>
<th>Grid</th>
<th>Passive</th>
<th>Active</th>
<th>Active</th>
<th>Active</th>
</tr>
</thead>
<tbody>
<tr>
<td>Grid Mode</td>
<td>n/a</td>
<td>Synchronous</td>
<td>Random</td>
<td>Random</td>
</tr>
<tr>
<td>(M) (cm)</td>
<td>10.16</td>
<td>11.43</td>
<td>11.43</td>
<td>11.43</td>
</tr>
<tr>
<td>(x/M)</td>
<td>70</td>
<td>62</td>
<td>62</td>
<td>62</td>
</tr>
<tr>
<td>(\nu) (m(^2) s(^{-1}))</td>
<td>15.7 \times 10^{-6}</td>
<td>16.0 \times 10^{-6}</td>
<td>16.0 \times 10^{-6}</td>
<td>16.0 \times 10^{-6}</td>
</tr>
<tr>
<td>(\beta) (°C/m)</td>
<td>7.4</td>
<td>15.6</td>
<td>2.7</td>
<td>3.6</td>
</tr>
<tr>
<td>(\kappa) (m(^2) s(^{-1}))</td>
<td>22.0 \times 10^{-6}</td>
<td>22.5 \times 10^{-6}</td>
<td>22.5 \times 10^{-6}</td>
<td>22.5 \times 10^{-6}</td>
</tr>
<tr>
<td>(U) (m s(^{-1}))</td>
<td>6.17</td>
<td>5.43</td>
<td>3.32</td>
<td>7.00</td>
</tr>
<tr>
<td>(\langle u^2 \rangle) (m(^2) s(^{-2}))</td>
<td>0.0219</td>
<td>0.121</td>
<td>0.0911</td>
<td>0.583</td>
</tr>
<tr>
<td>(\langle \theta^2 \rangle) (°C(^2))</td>
<td>0.147</td>
<td>3.65</td>
<td>0.800</td>
<td>582</td>
</tr>
<tr>
<td>(\langle \epsilon \rangle) ((= 15\nu \int_0^\infty k^2 F_{11}(k_1) dk_1)) (m(^2) s(^{-3}))</td>
<td>0.0521</td>
<td>0.278</td>
<td>0.0833</td>
<td>0.940</td>
</tr>
<tr>
<td>(\langle \epsilon_0 \rangle) (= \frac{3\kappa}{6\pi} \left\langle \left(\frac{du}{dx}\right)^2 \right\rangle) (°C(^2) s(^{-1}))</td>
<td>0.126</td>
<td>3.73</td>
<td>0.353</td>
<td>0.768</td>
</tr>
<tr>
<td>(R_A) (= \langle u^2 \rangle [15/\nu \epsilon]^{1/2})</td>
<td>94</td>
<td>222</td>
<td>306</td>
<td>582</td>
</tr>
<tr>
<td>(R_t) (= \langle u^2 \rangle^{1/2} [\nu \epsilon]^{1/2})</td>
<td>528</td>
<td>2960</td>
<td>5600</td>
<td>20300</td>
</tr>
<tr>
<td>(\eta) (= \left(\nu^3/\epsilon^{1/4}\right)) (m)</td>
<td>5.2 \times 10^{-4}</td>
<td>3.5 \times 10^{-4}</td>
<td>4.7 \times 10^{-4}</td>
<td>2.6 \times 10^{-4}</td>
</tr>
<tr>
<td>(\ell_u) (= \int_0^{\text{first zero}} \frac{\langle u(x + r)u(x)\rangle}{\langle u^2 \rangle}) (m)</td>
<td>0.069</td>
<td>0.12</td>
<td>0.30</td>
<td>0.39</td>
</tr>
<tr>
<td>(\ell_v) (= \int_0^{\text{first zero}} \frac{\langle v(x + r)v(x)\rangle}{\langle v^2 \rangle}) (m)</td>
<td>0.029</td>
<td>0.064</td>
<td>0.11</td>
<td>0.12</td>
</tr>
<tr>
<td>(\ell_\theta) (= \int_0^{\text{first zero}} \frac{\langle \theta(x + r)\theta(x)\rangle}{\langle \theta^2 \rangle}) (m)</td>
<td>0.057</td>
<td>0.16</td>
<td>0.34</td>
<td>0.44</td>
</tr>
<tr>
<td>(\ell') (= 0.9 [u^2]^{3/2}/\nu) (m)</td>
<td>0.056</td>
<td>0.14</td>
<td>0.30</td>
<td>0.43</td>
</tr>
<tr>
<td>(\ell'_\theta) (= \langle \theta^2 \rangle^{1/2}/\beta) (m)</td>
<td>0.052</td>
<td>0.12</td>
<td>0.33</td>
<td>0.29</td>
</tr>
</tbody>
</table>

**TABLE II.** Flow parameters.

### III. RESULTS AND DISCUSSION

The presentation of the results is subdivided into four sections. We begin in §3.1 with a derivation of an analogue to Hosokawa’s equation (Eq. 6) for the passive scalar field. In §3.2
we present incremental average structure functions for the scalar field, and compare them with those of the velocity field. In §3.3 we describe the probability density functions and conditional expectations of incremental averages. Lastly, we present a modified Yaglom’s equation, written for incremental sums (instead of incremental differences) in §3.4.

A. Passive Scalar Analogue of Hosokawa’s Equation

Before presenting an equation for the passive scalar field that is analogous to Hosokawa’s equation (Eq. 6) for the velocity field, a more detailed discussion of the derivation of the latter equation is in order. A general version of Hosokawa’s equation, which does not invoke the assumption of homogeneity of the field under consideration, can be derived. By (i) adding the (algebraic) expansions of \( \langle (\Delta u)^3 \rangle \) and \( \langle (\Delta u)(\Sigma u)^2 \rangle \), (ii) simplifying the expression by removing the terms that cancel out, and (iii) regrouping terms, one obtains the following expression:

\[
\langle (\Delta u)^3 \rangle + 4 \langle (\Delta u)(\Sigma u)^2 \rangle = 4 \langle (\Delta u)(\Sigma u^2) \rangle ,
\]  
(7)

which holds in all flows (independent of their degree of homogeneity, isotropy, etc.), given that the above result is purely algebraic.

Because the choice of variable in the above equation is arbitrary, it also can also be written for any scalar field:

\[
\langle (\Delta \theta)^3 \rangle + 4 \langle (\Delta \theta)(\Sigma \theta)^2 \rangle = 4 \langle (\Delta \theta)(\Sigma \theta^2) \rangle ,
\]  
(8)

This being said, a mixed-velocity-passive scalar form of this equation, which pertains to the scale-by-scale transfer of the scalar variance is more relevant to the present work. It can be derived in a manner analogous to that of Eq. (7), by adding the (algebraic) expansions of \( \langle (\Delta u)(\Delta \theta)^2 \rangle \) and \( \langle (\Delta u)(\Sigma \theta)^2 \rangle \). The result is:

\[
\langle (\Delta u)(\Delta \theta)^2 \rangle + 4 \langle (\Delta u)(\Sigma \theta)^2 \rangle = 4 \langle (\Delta u)(\Sigma \theta^2) \rangle ,
\]  
(9)

which, we reiterate, has not invoked the assumption of homogeneity. Experimental validation of this relation is shown in figure 1, where the two terms on the left side are shown as positive and the right side is negative such that the sum of the curves is zero.
Assuming homogeneity in the $x$-direction, equation (7) becomes equation (5), equation (8) becomes:

$$\langle (\Delta \theta)(\Sigma \theta)^2 \rangle = -\frac{1}{12} \langle (\Delta \theta)^3 \rangle,$$

and equation (9) becomes:

$$\langle (\Delta u)(\Sigma \theta)^2 \rangle = -\frac{1}{4} \langle (\Delta u)(\Delta \theta)^2 \rangle + \frac{1}{2} \langle u_{x+r}\theta_x^2 \rangle - \frac{1}{2} \langle u_x\theta_{x+r}^2 \rangle,$$

where $u_x = u(x)$, $u_{x+r} = u(x + r)$, and similarly for $\theta$.

Given the $\langle (\Delta u)(\Delta \theta)^2 \rangle$ term in the above equation, one can invoke Yaglom’s equation (in analogy with the invocation of Kolmogorov’s $4/5$ Law in Eq. 6), so that, in the inertial-convective subrange, equation (11) can be rewritten as:

$$\langle (\Delta u)(\Sigma \theta)^2 \rangle = \frac{1}{3} \langle \epsilon \theta \rangle r + \frac{1}{2} \langle u_{x+r}\theta_x^2 \rangle - \frac{1}{2} \langle u_x\theta_{x+r}^2 \rangle.$$

We note that equations (8) and (10) are not explicitly equations governing the (advec-tive) transport of the scalar field, given that they do not represent the advection of $\Sigma \theta$ (or $\Delta \theta$ for that matter). The advection of $\Sigma \theta$ (and its relation to that of $\Delta \theta$) is governed by equation (9). Furthermore, note the additional terms in equation (12), when compared to the analogous expression for the velocity field (equation (6)). These are not present in the analogous equation for the velocity field, as they cancel out when studying “velocity advecting velocity.” However, given the necessity of examining mixed velocity-scalar statistics when studying the transport of scalars [27, 29], the presence of these additional two terms becomes inevitable. Our last comment with respect to the above equations is that they serve as a guide to the types of statistics that will be examined in §III.B and §III.C.

To investigate Eq. (9), we plot the three mixed velocity-temperature structure functions (two mixed IASF/IDSFs and one IDSF) in that equation in figure 1. One observes that $4 \langle \Delta u \Sigma (\theta^2) \rangle$ balances $\langle (\Delta u)(\Sigma \theta)^2 \rangle$ at small scales (i.e., $r/\eta \lesssim 8$), whereas the balance is between $\langle (\Delta u)(\Delta \theta)^2 \rangle$ and $4 \langle (\Delta u)(\Sigma \theta)^2 \rangle$ for larger scales. And while IDSFs are influenced by the fluctuations at scales less than or equal to $r$, and IASFs are influenced by scales larger than or equal to $r$, mixed IASF/IDSFs, like $\langle (\Delta u)(\Sigma \theta)^2 \rangle$, uniquely depend on the length scale $r$. But given that the magnitude of fluctuations increases with the scale $r$, IDSFs like
FIG. 1. Scale-by-scale evolution of the terms in Eq. (9) at $R_\lambda = 222$. In (a), the terms are non-dimensionalized by $\langle \langle \epsilon_\theta \rangle \eta \rangle$, which is a constant. In (b), the terms are non-dimensionalized by $\frac{4}{3} \langle \epsilon_\theta r \rangle$, which is a non-dimensionalization that is both a function of scale, and analogous with Yaglom’s equation.

$\langle (\Delta u)(\Delta \theta)^2 \rangle$ obtain the majority of their contributions from scales close to $r$, and can effect the balance between an IDSF and a mixed IASF/IDSF.

B. Incremental Average Structure Functions (IASFs)

As an initial comparison of incremental averages of a passive scalar with those of the velocity field, we plot in figure 2(a) the second-order incremental average structure function of the of $u$, $v$, and $\theta$ for $R_\lambda = 582$, normalized by the variances of the respective quantities. One first observes that the present incremental averages of both $u$ and $v$ are similar to those observed by Mouri and Hori [21]; they decay from a value of 1 in the limit of $r \to 0$ to a value of 1/2, in the limit of $r \to \infty$, as expected. Furthermore, one also notes that (for separations smaller than the integral scale), $\langle (\Sigma_r v)^2 \rangle < \langle (\Sigma_r u)^2 \rangle$, a result that derives from the larger value of the longitudinal integral scale ($\ell_u$) as compared to the transverse one ($\ell_v$) in homogeneous, isotropic turbulence [30]. Given that $\ell_v < \ell_u$, $\langle (\Sigma_r v)^2 \rangle$ must decay from its small-scale limit (1) to its large-scale limit (1/2) over a shorter range of scales than $\langle (\Sigma_r u)^2 \rangle$. The tendency of $\langle (\Sigma_r v)^2 \rangle$ to values less than 1/2 at large scales (also observed for the boundary layer and jet data of Mouri and Hori [21], but, interestingly, not their grid turbulence data) can be attributed to the tendency of the autocorrelation of $v$ ($\rho_{vv}(r)$) to
FIG. 2. Normalized (a) second- and (b) fourth-order incremental averages of \( u, v \) and \( \theta \). \( R_\lambda = 582 \).

become negative at large scales to satisfy continuity (Tennekes and Lumley [30], p. 252-253).

With respect to \( \langle (\Sigma r \theta)^2 \rangle \), we observe that it is the slowest of the three functions to decay to its asymptotic value, which is consistent with the integral length scale of \( \theta \) (\( \ell_\theta \)) being larger than \( \ell_u \) and \( \ell_v \). (See Table II.) Fitting a curve of the form given in Table I to the second-order incremental averages results in best-fit scaling exponents, which are given in Table III. Note that these values must be the same as those for the second-order structure function, as is clear from the relations given in Table I.

<table>
<thead>
<tr>
<th>( R_\lambda )</th>
<th>94</th>
<th>306</th>
<th>582</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \gamma_u )</td>
<td>0.50</td>
<td>0.59</td>
<td>0.61</td>
</tr>
<tr>
<td>( \gamma_v )</td>
<td>0.43</td>
<td>0.55</td>
<td>0.57</td>
</tr>
<tr>
<td>( \gamma_\theta )</td>
<td>0.55</td>
<td>0.60</td>
<td>0.58</td>
</tr>
</tbody>
</table>

TABLE III. Scaling exponents of the curve fit of the form given in Table 1 for the second-order incremental average structure functions. Note that this equation renders these values the same as what would be obtained by measuring the scaling exponents of the second-order incremental difference structure functions. The ranges of \( r/\eta \) over which we fit these exponents varied with Reynolds number and quantity (i.e. \( u, v \) or \( \theta \)) and were calculated using the approach of [31].

It is also of interest to consider higher-order incremental averages. To this end, the fourth-order incremental average of \( u, v \), and \( \theta \) is plotted in figure 2(b) for \( R_\lambda = 582 \). (Note
that we do not plot third-order incremental averages, as these are effectively zero, due to the underlying symmetries of the velocity and temperature fields herein.) One observes that the fourth-order incremental averages are quite similar to those at second-order. They exhibit the same relative rates of decay towards their asymptotic values, dictated by the relative sizes of the integral length scales for each of the three fields. However, they instead asymptote to a value of approximately 0.25. Note that this latter value is consistent with the large-scale asymptote of $\langle (\Sigma \theta)^4 \rangle$, which, if one assumes homogeneity in the $x$-direction and the quasi-normal approximation to hold [32, 33], gives:

$$\lim_{r \to \infty} \langle (\Sigma \theta)^4 \rangle = \lim_{r \to \infty} \frac{1}{16} \langle (\theta(x+r) + \theta(x))^4 \rangle$$

$$= \frac{1}{16} \left[ \langle (\theta(x+r))^4 \rangle + 6 \langle (\theta(x+r))^2 \rangle \langle (\theta(x))^2 \rangle + \langle (\theta(x))^4 \rangle \right]$$

$$= \frac{1}{16} \left[ \langle (\theta(x+r))^4 \rangle + 6 \langle (\theta(x+r))^4 \rangle / 3 \right]^{1/2} \langle (\theta(x))^4 \rangle / 3 \right]^{1/2} + \langle (\theta(x))^4 \rangle \right]$$

$$= \frac{4}{16} \langle \theta^4 \rangle = \frac{\langle \theta^4 \rangle}{4}. \quad (13)$$

The Reynolds number dependence of the second and fourth-order incremental average structure functions of temperature are plotted in figure 3(a) and 3(b), respectively. The increase in the values of $r/\eta$ at which the structure function of the incremental average tend to their large-scale asymptotic values increases with Reynolds number, consistent with the increased separation of scales associated with larger Reynolds/Péclet numbers.

In figure 4, we proceed to consider the fourth-order incremental average of $u$, $v$, and $\theta$ non-dimensionalized by their respective second-order incremental averages (e.g., $\langle (\Sigma \theta)^4 \rangle / \langle (\Sigma \theta)^2 \rangle^2$), which we shall denote as “kurtosis structure functions of incremental averages.” As noted by Mouri and Hori [21], these must tend to the kurtosis of the respective quantities (e.g., $\langle \theta^4 \rangle / \langle \theta^2 \rangle^2$) as $r \to 0$, and to one half of the kurtosis of the respective quantities plus $\frac{\delta}{2}$ (e.g., $\langle \theta^4 \rangle / (2 \langle \theta^2 \rangle^2 + \frac{\delta}{2})$ as $r \to \infty$. Mouri and Hori [21] observed kurtosis structure functions of incremental averages that were constant to within $\pm 1\%$, but nevertheless exhibited very similar (small, but consistent) trends for $\Sigma u$ and $\Sigma v$ in grid turbulence, as well as for $\Sigma v$ in their turbulent boundary layer and jet. These were characterized by maxima at $r \sim \ell_u$ and minima at $r \sim 10^{-1} \ell_u$. They, however, noted that their observed kurtosis structure functions of $\Sigma u$ were considerably different due to the fact that the kurtosis of the $u$ velocity fluctuation was different from the Gaussian value of 3 (2.69 and 2.60
in the boundary layer and jet, respectively). In the present experiments, the variation of
the kurtosis structure functions of incremental averages is substantially larger than ±1% variation with $r/\eta$ observed by Mouri and Hori [21]. Although one might be tempted to
associate this increased variation with the intense, unsteady fluctuations associated with
active-grid generated turbulence, this larger variation of kurtosis structure functions of in-
cremental averages is also observed for the lower Reynolds number data set studied herein,
generated by means of a passive grid. In the present work, the kurtosis structure functions
of incremental averages of the $v$ component of velocity (figure 4(b)) exhibit the smallest
variations with $r/\eta$. Nevertheless, like in the work of Mouri and Hori [21], they too are
affected, at all scales, by the value of the kurtosis of $v$. Perhaps more interestingly, we
observe that the kurtosis structure function that exhibits the largest variation with $r/\eta$ is
that corresponding to temperature (figure 4(c)). We hypothesize that this larger variation of
$\langle (\Sigma \theta)^4 \rangle / \langle (\Sigma \theta)^2 \rangle^2$ with $r/\eta$ is associated with the intimate connection between the evolution
of passive scalar fields and the passive scalar initial conditions (Warhaft and Lumley [34],
Beaulac and Mydlarski [35]).

Given that turbulent transport of temperature fluctuations (i.e., the turbulent convective
heat transfer) is i) effected by the action of the turbulent velocity fluctuations against the
mean temperature gradient, and ii) a large-scale phenomenon, it is of interest to examine
the mixed (second-order) velocity-temperature IASFs: $\langle \Sigma v \Sigma \theta(r) \rangle / \langle v \theta \rangle$. We only consider
FIG. 4. Kurtosis structure functions of incremental averages. (a) $\langle (\Sigma u)^4 \rangle / \langle (\Sigma u)^2 \rangle^2$, $\langle (\Sigma v)^4 \rangle / \langle (\Sigma v)^2 \rangle^2$ and $\langle (\Sigma \theta)^4 \rangle / \langle (\Sigma \theta)^2 \rangle^2$ for $R_\lambda = 582$. (b) The Reynolds number dependence of $\langle (\Sigma v)^4 \rangle / \langle (\Sigma v)^2 \rangle^2$. (c) The Reynolds number dependence of $\langle (\Sigma \theta)^4 \rangle / \langle (\Sigma \theta)^2 \rangle^2$.

the quantity dependent upon the $v$-component on velocity, as that is the only one effecting net heat transfer given that the mean temperature gradient is solely in the $y$-direction. $\langle \Sigma v \Sigma \theta(r) \rangle / \langle v \theta \rangle$ is plotted in figure 5 for three different Reynolds numbers. One observes that these mixed velocity-temperature IASFs are similar to the respective second-order IASFs of their constituent components (i.e. $v$ and $\theta$). In general, it seems that IASFs of passive scalar fields behave similarly to those of the velocity field, even if the nature of the field is quite different (i.e. isotropic vs. anisotropic, with production vs. without production, etc.). This may be attributable to the small- and large-scale limits of IASFs, which are the same for all fields, and which therefore enforce a consistency in their form. However, the latter is invariably Reynolds- and Péclet-number-dependent, akin to what is also observed for IDSFs [31, 36, 37].

C. Probability Density Functions and Conditional Expectations of Incremental Averages

Given the contrast between the previously observed structure functions of incremental averages and those of differences, it is of interest to study the probability density functions (PDFs) of incremental averages, which are plotted in figure 6. As is well-known [9, 38], the PDF of incremental differences (e.g. $\Delta \theta$) changes with the scale under consideration, evolving from being quasi-Gaussian in shape at large separations, to being super-Gaussian for small increments, due to the effects of internal intermittency. On the other hand, one
observes that the PDF of $\Sigma \theta$ does not vary with $r/\eta$, as incremental differences are dominated by their large-scale behaviour [21]. As can be seen in figure 6(b), the PDF of $\Sigma \theta$ is nearly identical to the PDF of $\theta/\theta_{rms}$, consistent with the results of Sreenivasan and Dhruva [12] for the velocity field. Note that the PDF of $\theta/\theta_{rms}$ in this active-grid-generated flow is somewhat sub-Gaussian, due to the large integral length scale relative to the wind tunnel width – see Mydlarski and Warhaft [6].

We extend this analysis to joint probability density functions (JPDFs) of incremental averages of different quantities. To be able to sensibly interpret JPDFs of incremental averages, we first plot the (normalized) JPDFs of $u$ and $v$, $u$ and $\theta$, as well as $v$ and $\theta$ in figures 7 (a), (b) and (c), respectively, for $R_\lambda = 582$. We then proceed to plot (normalized) JPDFs of $\Sigma u$ and $\Sigma v$, $\Sigma u$ and $\Sigma \theta$, and $\Sigma v$ and $\Sigma \theta$ in figures 7(d), (e) and (f), respectively (for $r/\eta = 100$ and $R_\lambda = 582$). We remark that for a normalized joint-Gaussian distribution, circular contour lines indicate a lack of correlation between the two variables, and that statistical correlation is evidenced by contour lines that are elliptical with major axes parallel to i) $y = x$, for variables that are positively correlated, and ii) $y = -x$, for variables that are negatively correlated. This is most clearly demonstrated by the negative correlation between $v$ and $\theta$ in figure 7(c).

From figure 7, one observes that the JPDFs of the incremental averages are quite similar to their large-scale analogs (i.e. the JPDFs of $u$, $v$ and $\theta$). Furthermore, the relative insensitivity of JPDFs to the separation was confirmed by measurements performed at different
FIG. 6. Probability density function for the first order structure function of (a) the incremental difference and (b) the incremental average at $R_\lambda = 582$. Circles: $r = 10\eta$. Squares: $r = 100\eta$. Triangles: $r = 1000\eta$. Crosses: (a) PDF of $\frac{\partial \theta}{\partial x}/(\langle (\frac{\partial \theta}{\partial x})^2 \rangle)^{1/2}$ and (b) PDF($\theta/\theta_{rms}$). The solid lines are Gaussian fits to the (a) incremental difference at $r/\eta = 1000$ (mean: 0, standard deviation: 1.1) and (b) incremental average at $r/\eta = 1000$ (mean: -0.01, standard deviation: -1.2).

separations and Reynolds numbers (not shown), and is consistent with the PDFs of $\Sigma u$, $\Sigma v$ and $\Sigma \theta$, which were notably i) scale-independent, and ii) different from the PDFs of $\Delta u$, $\Delta v$ and $\Delta \theta$, which are strongly scale-dependent.

Conditional expectations of incremental averages are of further interest to examine, especially given their recent use in determining the effect of large-scale quantities on the small-scale structure of a flow (e.g. Kholmyansky and Tsinober [18], Blum et al. [19]). As previously noted, Mouri and Hori [21] calculated the average of the (squared) incremental average of the longitudinal velocity fluctuation conditioned upon its difference over the same separation ($\langle (\Sigma u)^2|\Delta u \rangle$) in grid turbulence, a boundary layer, and a turbulent jet to further investigate the (non-zero) correlation between $\Sigma u$ and $\Delta u$, given by equation (5). A related statistic, $\langle (\Delta u)^2|\Sigma u \rangle$, was calculated in Blum et al. [19] to investigate the effects of the large scales on the small-scale statistics in a variety of turbulent flows. Mouri and Hori [21] observed a clear difference in the behavior of $\langle (\Sigma u)^2|\Delta u \rangle$ in grid turbulence as compared to that observed in a boundary layer and a turbulent jet. This difference will be discussed shortly in relation to the analogous measurements of incremental averages in hydrodynamic.
R_{\lambda} = 582. \ r/\eta = 100.

and scalar fields.

In the present work, we examine \( \langle (\Sigma u_\alpha)^2 | \Delta u_\alpha \rangle \), \( \langle (\Sigma \theta)^2 | \Delta \theta \rangle \), and \( \langle (\Sigma \theta)^2 | \Delta u \rangle \) in figures 8 and 9. We observe good agreement between the present conditional expectations of \( \Sigma u \) and those of Mouri and Hori [21] measured in grid turbulence. Furthermore, note that the conditional expectations of \( \Sigma v \) follow a similar behaviour to those of \( \Sigma u \), albeit more symmetric, given the lack of any (known) relationship of the form of equation (5) existing for the transverse velocity fluctuation. The asymmetry in these plots is related to the non-zero correlation between the incremental averages and differences – see Mouri and Hori [21]. However, the expectations of \( \Sigma \theta \), conditioned upon \( \Delta \theta \), are distinctly different than the analogous ones for the velocity field. Whereas the latter are concave up, plots of \( \langle (\Sigma \theta)^2 | \Delta \theta \rangle \) are clearly concave down, indicating that large values of \( \Delta \theta \) are associated with small values of \( \Sigma \theta \), and vice versa. In fact, the present measurements of \( \langle (\Sigma \theta)^2 | \Delta \theta \rangle \) resemble the measurements of \( \langle (\Sigma u_\alpha)^2 | \Delta u_\alpha \rangle \) in the jet and boundary layers flows of Mouri and Hori [21], who hypothesized that their observed differences in \( \langle (\Sigma u_\alpha)^2 | \Delta u_\alpha \rangle \) for measurements in i) grid
turbulence, and ii) jet- and boundary-layer turbulence were due to the effects of turbulent production, which predominantly affect the largest scales. This explanation is consistent with the present scalar-field measurements, given that the temperature fluctuations in the present flow are produced by the turbulent velocity fluctuations acting on the mean scalar gradient. Consequently, the role of turbulent production (or presumably any other large-scale turbulent mechanism, such as decay or inhomogeneity) is important in the evolution of statistics of incremental averages, whether they be hydrodynamic or scalar. This being said, we remark that the dependence of \( \langle (\Sigma \theta)^2|\Delta \theta \rangle \) on the separation \( (r/\eta) \) in the present work (figure 8(b)) is weaker than that observed for \( \langle (\Sigma u_\alpha)^2|\Delta u_\alpha \rangle \) in the jet and boundary layer of Mouri and Hori [21]. Whereas the jet and boundary layer data of Mouri and Hori [21] exhibit a strong evolution with \( r/\eta \) over the range \( 10^1 < r/\eta < 10^3 \), the present results for \( \langle (\Sigma \theta)^2|\Delta \theta \rangle \) show a much weaker dependence on \( r/\eta \), with the shapes of \( \langle (\Sigma \theta)^2|\Delta \theta \rangle \) for all three separations \( (r/\eta = 10, 100 \text{ and } 1000) \) most closely resembling the results of Mouri and Hori [21] at \( r/\eta = 1000 \). This may be tied to the known connection between large and small scales of passive scalar fields (e.g. ramp-cliff structures[5, 10, 39]).

The Reynolds-number dependence of \( \langle (\Sigma u_\alpha)^2|\Delta u_\alpha \rangle \) and \( \langle (\Sigma \theta)^2|\Delta \theta \rangle \) is shown in figure 9. The dependence of \( \langle (\Sigma u_\alpha)^2|\Delta u_\alpha \rangle \) and \( \langle (\Sigma \theta)^2|\Delta \theta \rangle \) on \( R_\lambda \) is especially weak – a fact that is somewhat surprising given the (partially) large-scale nature of these statistics, which may not necessarily be flow-independent.
FIG. 9. Reynolds number dependence of the non-dimensionalized expectations of (a) $\langle (\Sigma u_\alpha)^2 \rangle |\Delta u_\alpha|^{1/2}$, (b) $\langle (\Sigma \theta)^2 \rangle |\Delta \theta|^{1/2}$, and (c) $\langle (\Sigma \theta)^2 \rangle |\Delta u_\alpha|^{1/2}$. In (a) and (c), the solid line represents data for which $u_\alpha = u$ (the longitudinal velocity fluctuation), and the dashed line represents $u_\alpha = v$ (the transverse velocity fluctuation). $R_\lambda = 94$ (bottom), $R_\lambda = 306$ (middle – offset by 0.5), and $R_\lambda = 582$ (top – offset by 1.0). $r/\eta = 10$.

D. Extension of Yaglom’s Equation

Given that $\langle (\Sigma \theta)^2 \rangle$ is a second-order statistic that is a function of the scale over which it is measured, it is of interest to consider its scale-by-scale budget. The objective of this section is thus to derive a scalar budget written in terms of the incremental average, analogous to Yaglom’s equation[40] (the scale-by-scale budget of the scalar variance, written in terms of the incremental difference).

Using the same procedure as outlined in Danaila and Mydlarski [22] for incremental differences, we write the advection-diffusion equation for the instantaneous scalar field $\theta_t$ at two points $\vec{x}$ and $\vec{x}^+$, which are separated by the increment $\vec{r} = \vec{x}^+ - \vec{x}$ and where the subscript ‘t’ denotes ‘total’:

$$\partial_t \theta_t + u_{i,t} \partial_i \theta_t = \kappa \partial_i^2 \theta_t,$$

and

$$\partial_t \theta_t^+ + u_{i,t}^+ \partial_i^+ \theta_t^+ = \kappa \partial_i^2 \theta_t^+.$$

The superscript $+$ refers to quantities evaluated at $\vec{x}^+$. In (14) and (15), $u_{i,t}$ is the instantaneous velocity vector, $\partial_i \equiv \partial / \partial t$, $\partial_i \equiv \partial / \partial x_i$ and $\partial_i^2$ is the Laplacian, $\partial^2 / \partial x_i^2$. Hereinafter, the notation $\partial_i$ and $\partial_i^+$ will be used to denote derivatives with respect to $x_i$ and $x_i^+$, respectively. When other spatial variables are involved, the derivatives will be written explicitly,
We assume that $u_{i,t}$ and $\theta_t$ depend upon $\vec{x}$ and not $\vec{x}^+$. Analogously, we also assume that $u_{i,t}^+$ and $\theta_t^+$ depend upon $\vec{x}^+$ and not $\vec{x}$. Furthermore, decomposing the total velocity field into mean and fluctuating components leads to $u_{i,t} = U_i + u_i$; $u_{i,t}^+ = U_i^+ + u_i^+$; $\theta_t = T + \theta$ and $\theta_t^+ = T^+ + \theta^+$.

Adding Eq. (14) to Eq. (15), then dividing by 2 yields an equation for the temperature IAFS ($\Sigma \theta$), viz.:

$$
\frac{D}{Dt}(\Sigma \theta) + \Sigma (U_i \partial_i T) + \Sigma (u_i \partial_i T) + u_i^+ \partial_i^+(\Sigma \theta) + u_i \partial_i(\Sigma \theta) =
\kappa(\partial_i^2 + \partial_i^{2+})(\Sigma \theta) + \kappa \partial_i^{2+}T^+ + \kappa \partial_i^2 T,
$$

where $\Sigma a \equiv (a + a^+)/2$ and $Da/Dt \equiv \partial_t a + U_i \cdot \partial_i a$, for an arbitrary field $a$. Moreover, note that $U_i$ and $T$ are, respectively, the mean velocity and temperature, for which statistical stationarity is assumed in both frames of reference, so that $\partial_t U_i = \partial_i U_i^+ = \partial_t T = \partial_i T^+ \equiv 0$. We also suppose that $\partial_i^2 T = \partial_i^{2+}T^+ \equiv 0$ (because $T$ is a linear function of $y$, and independent of $x$ and $z$), and that $\partial_t T = \partial_\alpha^+ T^+$ (the mean temperature gradient is the same at $\vec{x}$ and $\vec{x}^+$).

Following the approach suggested by Hill [41], we consider the gradient with respect to the midpoint of the interval:

$$
\vec{X} = \frac{1}{2}(\vec{x} + \vec{x}^+).
$$

such that:

$$
\partial_\alpha^+ \equiv \frac{\partial}{\partial r_\alpha} + \frac{1}{2} \partial_{X_\alpha};
$$

$$
\partial_\alpha \equiv -\frac{\partial}{\partial r_\alpha} + \frac{1}{2} \partial_{X_\alpha},
$$

resulting in $\partial_{X_\alpha} = \partial_\alpha + \partial_\alpha^+$.

By taking into account (18), multiplying Eq. (16) by $2\Sigma \theta$, and averaging we finally obtain

$$
D_i\langle(\Sigma \theta)^2\rangle(\vec{r}) + 2\langle\Sigma u_i \Sigma \theta\rangle \partial_i T(\vec{r}) + \frac{1}{2} \langle[\partial_i + \partial_i^+] \cdot [u_i + u_i^+] (\Sigma \theta)^2\rangle(\vec{r})
$$

$$
+ \frac{\partial}{\partial r_i} \langle\Delta u_i (\Sigma \theta)^2\rangle(\vec{r}) = 2\kappa \frac{\partial^2}{\partial r_i^2} \langle(\Sigma \theta)^2\rangle(\vec{r}) - \frac{1}{2} \langle(\epsilon_\theta) + (\epsilon_\theta)^+\rangle.
$$

Note that the first line in equation (20) corresponds to large-scale effects. More specifically, the first term represents the non-stationarity and/or advection by way of the mean velocity, the second one is a production term, whereas the third one pertains to the turbulent
diffusion. In (20), each term depends on the spatial vector $\vec{r}$. Special attention should be paid to the last term, e.g. $\langle \epsilon_\theta \rangle + \langle \epsilon_\theta \rangle^+$ which also depends on the vector $\vec{r}$. We now proceed to write Eq. (20) as:

$$D + P + T d + \frac{\partial}{\partial r_i} \langle \Delta u_i (\Sigma \theta)^2 \rangle (\vec{r}) = 2\kappa \frac{\partial^2}{\partial r_i^2} \langle (\Sigma \theta)^2 \rangle (\vec{r}) - \frac{1}{2} (\langle \epsilon_\theta \rangle + \langle \epsilon_\theta \rangle^+),$$

(21)

where the first term on the LHS of (21) is the decay term ($D$), the second one is the production term ($P$), and the third one is the turbulent diffusion term, $Td$. All the others are the classical terms, analogous to those in Yaglom’s equation, that will be therefore treated in the classical manner [22]. Also note that both the small- and large-scale limits of Eq. (21) are consistent with the one-point energy budget equation. Moreover, recall that the large-scale limit ($r \to \infty$) the scale-by-scale budget equation simplifies to the one-point energy (scalar variance) budget, whereas the small-scale limit reduces to the evolution equation of the mean dissipation rate of the scalar variance, $\langle \epsilon_\theta \rangle$. Therefore, the requirement that equation (21) be balanced at small scales imposes a stronger constraint – a discussion to which we will return in the context of the experimentally-obtained results.

We further note, in the context of the locally stationary, homogeneous (region of the) flow, that the flow is statistically stationary at a given position, so that time derivatives of any time-averaged statistics are zero. The first term is therefore equal to $U_i \partial_i \langle (\Sigma \theta)^2 \rangle (\vec{r})$.

Terms in equation (21) can be evaluated from either direct numerical simulations, or combined planar velocity and scalar measurements (e.g. Planar Laser Induced Fluorescence (PLIF) combined with Particle Image Velocimetry (PIV)). In the present work, we evaluate terms in equation (21) from (one-point) simultaneous hot- and cold-wire measurements. This approach does not allow us to estimate the real, spatial variation of these terms (without performing concurrent measurements at two points in space, thus involving two X-wires and two cold-wires operating simultaneously). Given this context, our approach involves assuming all extra terms take on an isotropic form. Their estimation from experimental (hot- and cold-wire) data implicitly assumes that the two reference systems are actually identical, and the spatial increments and/or IAFS at a scale $r$ are estimated using Taylor’s hypothesis, which imposes homogeneity at all scales.

For relatively small scales, the molecular diffusion and the advection terms are considered as being locally isotropic and therefore only depend on $r$ (the modulus of the separation $\vec{r}$). Thus, the divergence and Laplacian operators assume particular forms under these
conditions. Note that the hypothesis of local isotropy of $\langle (\Sigma \theta)^2 \rangle$ is less likely to be accurate than the same hypothesis for $\langle (\Delta \theta)^2 \rangle$ because the former includes all scales larger than a given scale $r$ (which are more likely anisotropic), whereas the latter pertains to the scales smaller than or equal to $r$.

For increasingly large scales, different terms in Eq. (21) tend to constants, and their spatial dependence is no longer observed, as discussed in [22]. Therefore, application of local isotropy in our present approach does not ignore the anisotropy of the flow, and our approach is not overly sensitive to this assumption.

Therefore, under the assumption of local isotropy, the terms $D$, $P$ and $T d$ can be considered as independent of the spatial orientation. Our last hypothesis is that the variations in $\langle \epsilon_\theta \rangle$ over scales smaller than $L$ (the integral scale) are negligible so that $\langle \epsilon_\theta \rangle = \langle \epsilon_\theta \rangle^+$. Thus, the term containing the two-point mean dissipation rate: $\langle \epsilon \rangle + \langle \epsilon \rangle^+$, does not depend on the spatial orientation ($\vec{r}$).

Since all the terms in Eq. (21) now only depend on $r$, suitable expressions for the divergence and Laplacian operators can be chosen (namely those corresponding to a spherical coordinate system). By following the classical approach [22], we finally obtain

$$-\langle \Delta u_1 (\Sigma \theta)^2 \rangle + 2\kappa \frac{d}{dr} \langle (\Sigma \theta)^2 \rangle - \frac{U_1}{r^2} \int_0^r s^2 \partial_1 \langle (\Sigma \theta)^2 \rangle ds$$

$$- \frac{2}{r^2} \int_0^r s^2 \left[ \frac{\partial T}{\partial x_2} \langle \Sigma u_2 \Sigma \theta \rangle + \partial_2 \langle (\Sigma u_2) (\Sigma \theta)^2 \rangle \right] ds = \frac{1}{3} \langle \epsilon_\theta \rangle r,$$

where $s$ is a dummy variable. As shown by Danaila and Mydlarski [22], turbulent diffusion is negligible in the flow under investigation at the level of one-point energy budget equation, which represents the large-scale limit of Eq. (22). The final transport equation for $\langle (\Sigma \theta)^2 \rangle$ is:

$$-\langle \Delta u_1 (\Sigma \theta)^2 \rangle + 2\kappa \frac{d}{dr} \langle (\Sigma \theta)^2 \rangle - \frac{U_1}{r^2} \int_0^r s^2 \partial_1 \langle (\Sigma \theta)^2 \rangle ds$$

$$- \frac{2}{r^2} \frac{\partial T}{\partial x_2} \int_0^r s^2 [\langle \Sigma u_2 \Sigma \theta \rangle] ds = \frac{1}{3} \langle \epsilon_\theta \rangle r,$$

At this point, a reasonable assumption can be made about the character of the second-order moment of the scalar incremental difference. Following the approach of Danaila et al. [42], we assume a self-similar behavior of the downstream evolution of structure functions such that they can be written as:

$$\langle (\Sigma \theta)^2 \rangle (r, x_1) = f(r) g(x_1).$$  \hspace{1cm} (24)
Suitable choices for \( f(r) \) and \( g(x_1) \) are:

\[
 f(r) = \frac{\langle (\Sigma \theta)^2 \rangle}{\langle \theta^2 \rangle} = \text{constant}
\]

\[
 g(x_1) = \langle \theta^2 \rangle
\]

Implementing this assumption, the final form of the transport equation for \( \langle (\Sigma \theta)^2 \rangle \) becomes:

\[
\begin{align*}
 -\langle \Delta u_1 (\Sigma \theta)^2 \rangle + 2\kappa \frac{d}{dr} \langle (\Sigma \theta)^2 \rangle - \frac{U_1}{r^2} \partial_1 \langle \theta^2 \rangle \int_0^r s^2 \frac{\langle (\Sigma \theta)^2 \rangle}{\langle \theta^2 \rangle} ds \\
 - \frac{2}{r^2} \partial_{x_2} \langle u_2 \theta \rangle \int_0^r s^2 \left[ \frac{\langle \Sigma u_2 \Sigma \theta \rangle}{\langle u_2 \theta \rangle} \right] ds = \frac{1}{3} \langle \epsilon_\theta \rangle r,
\end{align*}
\]

(25)

After dividing by \( \langle \epsilon_\theta \rangle r \) and multiplying by 3, this equation can be written in the following dimensionless form:

\[
 A^* + B^* + D^* + P^* = 1,
\]

(26)

where \( A^* \) is the (non-dimensional) advection term, \( B^* \) the molecular one, \( D^* \) is the inhomogeneous ('decay') term in the streamwise direction \( x_1 \), and \( P^* \) is the production term.

This scalar budget, corresponding to equation (26), is plotted in figure 10. (Note that both downstream and transverse measurements were made, such that quantities like \( \partial \langle \theta^2 \rangle / \partial x_1 \) and \( \partial T / \partial x_2 \) were evaluated from experimental data.) Taking into consideration that \( \langle (\Sigma \theta)^2 \rangle \) represents the total scalar variance at all scales larger or equal to \( r \), the physical significance can be considered. Specifically, the total scalar variance transferred towards smaller scales by all scales \( \geq r \), i.e. term proportional to \( \langle \epsilon_\theta \rangle \), is composed of four types of contributions:

(i) positive contribution due to the variance produced in the flow at scales larger than or equal to \( r \) (term \( P^* \))

(ii) negative contribution due to the molecular diffusion (term \( B^* \))

(iii) negative contribution due to the ‘decay’ effect (term \( D^* \)). Recall that the scalar variance increases in the downstream direction in this flow, unlike for the velocity field grid turbulence. This is due to the presence of the mean scalar scalar gradient, which continuously injects scalar fluctuations in the flow.
(iv) negative contribution over the majority range of scales (term $A^*$), which correspond to the energy transferred through turbulence. However, for the smallest scales (here, smaller than $8\eta$), this term changes its sign and its value is $-\langle \Delta u (\Delta \theta)^2 \rangle$. The physical signification of term $A^*$ is directly related to the balance between $-\langle \Delta u (\Delta \theta)^2 \rangle$ and $\langle \Delta u \Sigma (\theta^2) \rangle$, as reflected by Eq. (9) and Fig. 1.

![FIG. 10. Scale-by-scale evolution of the terms in equation (26). $R_\lambda = 222$.](image)

Finally, we both note and explain the origin of the disagreement between Eq. (26) and the experimental data at small scales. Scale-by-scale equations for incremental differences, by definition, reduce to the definition of the dissipation rate of turbulent kinetic energy, or scalar variance, which is well validated experimentally [22]. In the present case, as already stated and in contrast with the scale-by-scale equations for incremental differences, the small-scale limit of Eq. (26) reduces to the sum of the:

(i) definition of the dissipation rate of turbulent kinetic energy, or scalar variance, and
(ii) one-point turbulent scalar variance budget, which has contributions from all scales.

One might therefore hypothesize that this disagreement between Eq. (26) and the experimental data at small scales is due to a combined effect of:

(i) residual anisotropy, and

(ii) a loss of accuracy in an assumption made when deriving Eq. (26).

Specifically, the two measurements in our incremental average are no longer statistically independent at small separations (i.e., as \( \vec{r} \) tends to zero), which conflicts with the assumption that the two points \( \vec{x} \) and \( \vec{x}^+ \) are independent (i.e., \( u_{i,t} \) and \( \theta_t \) depend only on \( \vec{x} \), and \( u_{i,t}^+ \) and \( \theta_t^+ \) depend only on \( \vec{x}^+ \)) – an assumption used in the both the derivation of Eq. (26), as well as that of the scale-by-scale equations for incremental differences. When \( \vec{r} = 0 \), this assumption fails completely because the two points are coincident (identical). This unique phenomenon does not arise when considering incremental differences, because the averages cancel out and the scale-by-scale evolution equation of incremental differences reduces to the evolution equation of \( \langle \epsilon \rangle \) or \( \langle \epsilon_\theta \rangle \). When considering incremental sums rather than differences, the averages do not cancel and are therefore correlated at small scales, in contradiction with the assumption underlying the derivation of Eq. (26).

If one does not make the hypothesis that the two quantities are independent in deriving Eq. (26), one obtains:

\[
A^* + M^* + D^* + P^* = 1,
\]

where the molecular term \( M^* = 3M/\langle \epsilon_\theta \rangle r \) (which replaces the term \( B^* \) in Eq. (26)) and \( M \) is given by:

\[
M = \frac{1}{r^2} \int_0^r s^2 \left[ \left\langle \frac{\partial \theta}{\partial x_j} \frac{\partial \theta^+}{\partial x^+_j} \right\rangle + \left\langle \frac{\partial \theta^+}{\partial x^+_j} \frac{\partial \theta}{\partial x_j} \right\rangle \right] ds = \frac{2}{r^2} \int_0^r s^2 \left[ \left\langle \frac{\partial \theta}{\partial x_j} \frac{\partial \theta^+}{\partial x^+_j} \right\rangle \right] ds. \quad (28)
\]

When the two points are statistically independent, the term \( \left\langle \frac{\partial \theta}{\partial x_j} \frac{\partial \theta^+}{\partial x^+_j} \right\rangle \) is equal to zero. However, when \( r \to 0 \), this term becomes proportional to the dissipation rate of the scalar variance and balances the budget equation. The lack of statistical independence at small scales could also contaminate the advection term \( A^* \) and require a similar analysis, although we do not do so herein, because the advection term is present at all scales, and for
most of them the two frames are indeed independent. The correlation between two points at small separations offers a new perspective on the statistical approach to understanding turbulent quantities at a scale $r$ such as $\Sigma \theta$ or any other quantity that are more complex than classical incremental differences.

IV. CONCLUSIONS

In conclusion, we have investigated the statistics of incremental averages of a turbulent passive scalar for the first time. We studied both their spatial (i.e. $r/\eta$) and Reynolds number dependence, as well as certain relevant mixed (velocity-temperature) statistics of incremental averages. In doing so, we derived a passive scalar analogue to Hosokawa’s equation.

Although statistics of incremental averages are \textit{primarily} large-scale quantities, and thus expected to be flow-dependent, the statistics of incremental averages of passive scalars measured herein exhibit many similarities to those of the velocity field in different flows. Nevertheless, the similarities are not universal and certain statistics exhibit distinct differences from the velocity field. Furthermore, although IASFs are dictated by the scales larger than or equal to $r$, it is predominantly the largest scales that contribute the to their behaviour, as was clearly demonstrated by the PDFs and JPDFs. Conditional expectations of $\Sigma \theta$, conditioned upon $\Delta \theta$, were found to be distinctly different than the analogous ones for the velocity field, and more closely resembled the conditional expectations $\langle (\Sigma u_{\alpha})^2 | \Delta u_{\alpha} \rangle$ measured in the jet and boundary layers (i.e. shear) flows of Mouri and Hori [21], and may be tied to the presence of mean (i.e. large-scale) gradients of temperature and velocity, respectively.

We also derived a scale-by-scale evolution equation for the incremental average of passive scalars, similar to Yaglom’s equation. Agreement of the data was reasonable, given the assumptions involved in its derivation, and those used in the analysis of our (one-point) data.

Future studies of incremental averages may want to further study and quantify the effect of the assumption of statistical independence of quantities measured at $\vec{x}$ and $\vec{x}^+$ as $\vec{r} \to 0$. Moreover, it may also be of interest to examine the effect of the nature of the large-scales of the (scalar) field (i.e. its generation mechanism) on the structure and evolution IASFs. For example, scale-by-scale budgets of $\langle (\Sigma \theta)^2 \rangle$ could be derived for flows in which the large
scales are inhomogeneous, as opposed to the homogeneous flow studied herein. Such an extension of this work may be fruitful, given the i) similarities observed in the statistics of \( \langle (\Sigma \theta)^2 \rangle \) presented herein, as well as those of \( \langle (\Sigma u_\alpha)^2 \rangle \) presented by Mouri and Hori [21] in jet and boundary layer flows — flows in which fluctuations are produced by mean gradients, and ii) differences with the velocity IASFs measured in homogeneous, isotropic grid turbulence, both herein and by Mouri and Hori [21]. Such would could, hopefully, indicate whether any possibility exists that IASFs may exhibit a potential universal behavior, or whether they are invariably Reynolds/Péclet-number dependent, given their contribution from scales greater than or equal to \( r \).

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