FEDSM2000-11197

VISUALIZATION OF SHEAR STRESS WITH MICRO IMAGING CHIP AND DISCRETE WAVELETS TRANSFORM

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Keywords: Discrete Wavelets Transform, Frequency Analysis, Turbulent Flow, Boundary Layer, MEMS, Shear Stress Sensor

ABSTRACT

Two dimensional stripe structure in turbulent boundary layer has been clearly visualized by a combination of a shear stress sensor using MEMS (Micro-Electro-Mechanical-Systems) and discrete wavelets transform. The MEMS shear stress micro chip is designed and fabricated by surface micro-machining technology, contributing to obtaining the time-space two dimensional shear stress data. The discrete wavelets transform is a software technique to decompose the frequency level with the time and space information of the wave form. The experiments for the shear stress distribution were carried out on Re = 6960, 12180 and 17400. This technique for a single phase we is applicable to multiphase flow analysis.

INTRODUCTION

Our research group has been studying a spiral multiphase flow which is a swirling flow with a large free vortex region, a high concentration to the axis and high stability [Horii 1990]. From the characteristics, the spiral flow is useful for a high performance pneumatic transportation without particles touching pipe inner wall due to the stable eddy structure [Takei 1997]. Accordingly, it is necessary to analyze near-wall shear stress of the solid-air two phase flow to improve the system because the particle stability in solid-air pipeline is affected by eddy structure due to shear stress which occurs near the wall. The presence of near-wall shear stress streaks in turbulent boundary layer has been observed for many years in flow visualization and experimental investigation [Cantwell 1891,

Kim H. T 1971, Falco 1980, Head 1981, Smith 1983]. At high Reynolds numbers, these streaks are typically very small in size and cannot be properly resolved by traditional measuring techniques. Numerical simulations indicate that the streaks are associated with streamwise vortices in the viscous sub-layer. The rotational motion of these vortices imposes high fluctuating surface shear stress on the wall [Kim, J. 1987]. There have been many measurement techniques for measuring shear stress. The hot-film technique and its variants have been widely used for the detailed investigations of fluctuating wall shear stress [Alfredsson 1988, Bruun 1995]. The direction sensitive laser Doppler anemometer is a candidate, which enables to evaluate both magnitude and direction of the wall shear stress. An optical method proposed by Naquwi is supposed to be capable of measuring the wall shear stress with high spatial resolution [Naquwi and Reynolds 1991]. No matter the instrument, the requirements of fine spatial resolution, fast frequency response, high sensitivity and convenience need to be satisfied for turbulent boundary layer

Recently, the availability of a new manufacturing process, micro-electro-mechanical-systems (MEMS) technology, has offered the possibility of sensing and controlling the small nearwall streaks [Ho 1997]. A multidisciplinary research collaboration between UCLA and Caltech has undergone to design and fabricate a large-scale distributed control system with integrated micromachined transducers and microelectronic

circuits for surface shear stress control in turbulent boundary layers [Tung 1995, Ho 1996].

However, the distribution of stripe structure with MEMS is the integral value that is composed of various kinds of frequency ingredients. It leads to make the stripe structure ambiguous. Therefore, new techniques such as statistics analysis and frequency analysis to the image data is necessary for analyzing the structure in details. The stripe structure obtained with the micro imaging chip is clearly visualized with the statistics method [Kimura 1997]. In terms of the frequency analysis, Fourier transform is a popular method; however, the transform removes the time and space information of the stripe structure. Currently, wavelets transform has been started using for timespace frequency analysis instead of Fourier transform in mechanical engineering fields. The merits of the wavelets analysis is to be able to analyze the frequency not to erase the time-space information. Wavelets transform [Moret 1989] is roughly classified with two types, which are continuous wavelets transform and discrete wavelets transform.

ntinuous wavelets transform has been generally used for time requency analysis in vibration wave. For example, self-similarity of a jet inner structure [Everson 1990], breakdown of a large eddy and successive branching in a plane jet [Li 1995], decomposition of Reynolds stress in a jet [Gordeyev 1995], and multiple acoustic modes and shear layer instability [Walker 1995] were investigated. However, most of the researchers on the time-frequency analysis carried out the continuous wavelets transform. On the other hand, the discrete wavelets transform has been mainly used for picture image processing. The analysis enables to decompose and to compose picture image data quantitatively because of the orthonormal transform. Saito applied this idea to analyzing electromagnetic wave[Saito 1996].

The originality of this paper lies in applying discrete wavelets transform to visualizing the wall shear stress obtained with micro shear stress imaging chip. In this study, the stripe structure of shear stress in a turbulence boundary layer are extracted on various frequency levels with discrete wavelets nsform.

THEORY OF DISCRETE WAVELETS TRANSFORM Basic Concept Using Simple Base Function

Basic concept of discrete wavelets transform is described using matrix expression instead of integral expression. One dimensional input data matrix with four elements **X** and an analyzing wavelets matrix of Haar base function **W** are used to simplify the expression. For example, the input data matrix **X** is one dimensional shear stress data. The wavelets transform matrix **S** that indicates wavelets spectrum is expressed by

$$\begin{bmatrix} S_1 \\ D_1 \\ d_1 \\ d_2 \end{bmatrix} = \left(\frac{1}{\sqrt{2}}\right)^2 \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ \sqrt{2} & -\sqrt{2} & 0 & 0 \\ 0 & 0 & \sqrt{2} & -\sqrt{2} \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \qquad \qquad (1)$$

or
$$S = W \cdot X$$
 —(2)

Where, $W^T \cdot W = I$, I is a unit matrix and W^T is a transpose matrix of W. The analyzing wavelets matrix is an orthonormal. In Eq. (1), the first element in the wavelets spectrum S_I shows a transform to obtain a mean value with a weight on the all input data, a+b+c+d. The second element in the wavelets spectrum D_1 shows a transform to obtain a difference value between the fast half and the latter half with a weight on the input data, $\{(a+b)-(c+d)\}$. It means that this element includes the lower frequency level of the input data. The third element d_I shows a transform to obtain a difference value on the first half of the input data, (a-b). The fourth element d_2 is a transform to obtain a difference value on the latter half, (c-d). The third and forth elements include the higher frequency level of the input data. Therefore, the input data is able to classified to a range from higher frequency level to lower frequency level. Because of the orthonormal function, the inverse discrete wavelets transform is expressed by,

$$\mathbf{X} = \mathbf{W}^{\mathbf{T}} \cdot \mathbf{S} \qquad \qquad (3)$$

Moreover, from Eq.(3), the input data X is decomposed by multiresolution. The matrix expression is

$$\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{pmatrix} 1 \\ \frac{1}{\sqrt{2}} \end{pmatrix}^{2} \begin{bmatrix} 1 & 1 & \sqrt{2} & 0 \\ 1 & 1 & -\sqrt{2} & 0 \\ 1 & -1 & 0 & \sqrt{2} \\ 1 & -1 & 0 & -\sqrt{2} \end{bmatrix} \begin{bmatrix} S_{1} \\ D_{1} \\ d_{1} \\ d_{2} \end{bmatrix} = \mathbf{W}^{\mathsf{T}} \begin{bmatrix} S_{1} \\ 0 \\ 0 \\ 0 \end{bmatrix} + \mathbf{W}^{\mathsf{T}} \begin{bmatrix} 0 \\ D_{1} \\ 0 \\ 0 \end{bmatrix} + \mathbf{W}^{\mathsf{T}} \begin{bmatrix} 0 \\ 0 \\ d_{1} \\ d_{2} \end{bmatrix}$$
or $\mathbf{X} = \mathbf{W}^{\mathsf{T}} \mathbf{S} = \mathbf{W}^{\mathsf{T}} \mathbf{S}_{0} + \mathbf{W}^{\mathsf{T}} \mathbf{S}_{1} + \mathbf{W}^{\mathsf{T}} \mathbf{S}_{2}$

$$\mathbf{W} \text{here, } \mathbf{S}_{0} = [S_{1} \ 0 \ 0 \ 0]^{T} \quad \mathbf{S}_{1} = [0 \ D_{1} \ 0 \ 0]^{T} \quad \mathbf{S}_{2} = [0 \ 0 \ d_{1} \ d_{2}]^{T}.$$

$$\mathbf{W}^{\mathsf{T}} \mathbf{S}_{0} , \mathbf{W}^{\mathsf{T}} \mathbf{S}_{1} \text{ and } \mathbf{W}^{\mathsf{T}} \mathbf{S}_{2} \text{ are called Level 0, Level 1 and Level 2, respectively.}$$

Generalization of Discrete Wavelets Transform

Many orthonormal wavelets analyzing functions are found [Molet 1989]. The basic concept of the discrete wavelets transform is generalized by using fourth Daubechies function. The analyzing wavelets matrix is also an orthonormal function. The analyzing wavelets matrix **W** is acquired by a cascade algorithm on the basis of a function matrix **C**. The matrix **C** is shown in Eq.(6),

$$\mathbf{C} = \begin{pmatrix} c_0 & c_1 & c_2 & c_3 & 0 & 0 & \cdot & 0 & 0 & 0 & 0 \\ c_3 & -c_2 & c_1 & -c_0 & 0 & 0 & \cdot & 0 & 0 & 0 & 0 \\ 0 & 0 & c_0 & c_1 & c_2 & c_3 & \cdot & 0 & 0 & 0 & 0 \\ 0 & 0 & c_3 & -c_2 & c_1 & -c_0 & \cdot & 0 & 0 & 0 & 0 \\ \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & 0 & \cdot & c_0 & c_1 & c_2 & c_3 \\ 0 & 0 & 0 & 0 & 0 & 0 & \cdot & c_3 & -c_2 & c_1 & -c_0 \\ c_2 & c_3 & 0 & 0 & 0 & 0 & \cdot & 0 & 0 & c_0 & c_1 \\ c_1 & -c_0 & 0 & 0 & 0 & 0 & \cdot & 0 & 0 & c_3 & -c_2 \end{pmatrix} c_3 = \frac{1 + \sqrt{3}}{4\sqrt{2}}$$

 $c_3 - c_2 + c_1 - c_0 = 0$ —(8) Where, $\mathbf{C}^T \cdot \mathbf{C} = \mathbf{I}$. The first line in Eq.(6) is called scaling coefficients and second line is called wavelets coefficients. Forth Daubechies function has four coefficients in a line. The first line shows a transform to obtain a mean value with weights of c_0 , c_1 , c_2 and c_3 on the input data. The second line shows a

transform to obtain a difference value with weights of c_0 , c_1 , c_2 and c_3 on the input data. The third line shows a transform to translate the first line by two steps. The fourth line is a transform to do the second line by two steps. Eqs.(7) and (8) show the transformed values are zero when the input data are constant or are simply increased. To explain easily the process to acquire the analyzing wavelets matrix W from C, the matrix C is assumed to be one dimensional 16 elements,

$$\mathbf{X} = \left[x_1 \, x_2 \, x_3 \, x_4 \, x_5 \, x_6 \, x_7 \, x_8 \, x_9 \, x_{10} \, x_{11} \, x_{12} \, x_{13} \, x_{14} \, x_{15} \, x_{16} \right]^T \qquad ----(9)$$

From Eqs. (6) and (9), the transformed matrix X' is

$$\mathbf{X'} = \mathbf{C}_{16}\mathbf{X} = [\mathbf{S}_1 \, d_1 \, \mathbf{S}_2 \, d_2 \, \mathbf{S}_3 \, d_3 \, \mathbf{S}_4 \, d_4 \, \mathbf{S}_5 \, d_5 \, \mathbf{S}_6 \, d_6 \, \mathbf{S}_7 \, d_7 \, \mathbf{S}_8 \, d_8]^T \quad ----(10)$$

Where, C_{16} is 16X16 matrix of C. The element s indicates the mean value and the element d indicates the difference value. The elements in the matrix X' are replaced by a matrix P_{16} .

 $P_{16}X' = P_{16} C_{16}X = [s_1 s_2 s_3 s_4 s_5 s_6 s_7 s_8 d_1 d_2 d_3 d_4 d_5 d_6 d_7 d_8]^T$ —(11) Where, P_{16} is defined as

Moreover, from Eq.(11), the transform is continuously carried out by **C** and **P** without any operations to the difference values, $\mathbf{W^{(2)}X} = [S_1 S_2 S_3 S_4 D_1 D_2 D_3 D_4 d_1 d_2 d_3 d_4 d_5 d_6 d_7 d_8]^T - (13)$ $\mathbf{S} = \mathbf{W^{(3)}X} = [S_1 S_2 D_1 D_2 D_1 D_2 D_3 D_4 d_1 d_2 d_3 d_4 d_5 d_6 d_7 d_8]^T - (14)$ Where,

$$W^{(2)} = (P_{16}'C'_{16})(P_{16}C_{16}) \qquad (15)$$

$$W^{(3)} = (P_{16}'' C_{16}'')(P_{16}' C_{16}')(P_{16}C_{16}) \qquad (16)$$

$$P_{16}' = \begin{bmatrix} P_{8} & 0 \\ 0 & I_{8} \end{bmatrix} C_{16}' = \begin{bmatrix} C_{8} & 0 \\ 0 & I_{8} \end{bmatrix} P_{16}'' = \begin{bmatrix} P_{4} & 0 \\ 0 & I_{12} \end{bmatrix} C_{16}'' = \begin{bmatrix} C_{4} & 0 \\ 0 & I_{12} \end{bmatrix} \qquad (17)$$

 $W^{(3)}$ is a analyzing wavelets matrix that is W in Eq.(2). The vavelets spectrum S in Eq.(2) is $W^{(3)}X$ in Eq. (14). In Eq.(13), S_I indicates the mean value from s_I to s_I in Eq.(11). S_2 indicates the mean value from s_I to s_I in Eq.(11). S_I indicates the difference value from s_I to s_I . In Eq.(14), S_I indicates the mean value from S_I to S_I in Eq.(13). D_I indicates the difference value from S_I to S_I in Eq.(13). From Eq.(14), the input data are transformed to the mean values and the difference values with valuable resolution levels by the discrete wavelets transform. The input data are divided into a range from high frequency to low frequency.

From Eq.(14), the inverse wavelets transform is,

$$X = [W^{(3)}]^{T}S \qquad ----(18)$$

$$[W^{(3)}]^{T} = [(P_{16}"C_{16}")(P_{16}'C_{16}")(P_{16}C_{16})]^{T}$$

$$= C_{16}^{T}P_{16}^{T}(C_{16}")^{T}(P_{16}")^{T}(C_{16}")^{T}(P_{16}")^{T} ----(19)$$

From Eq.(18), the multiresolution is,

$$\mathbf{X} = [\mathbf{W}^{(3)}]^{\mathrm{T}} \mathbf{S} = [\mathbf{W}^{(3)}]^{\mathrm{T}} \mathbf{S}_0 + [\mathbf{W}^{(3)}]^{\mathrm{T}} \mathbf{S}_1 + [\mathbf{W}^{(3)}]^{\mathrm{T}} \mathbf{S}_2 + [\mathbf{W}^{(3)}]^{\mathrm{T}} \mathbf{S}_3 - \dots - (20)$$

Where,

$$S_0 = [S_1 S_2 0 0 0 0 0 0 0 0 0 0 0 0]^T$$

$$S_1 = [0 \ 0 \ D_1 \ D_2 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]^T$$

$$\mathbf{S_2} = [0\ 0\ 0\ 0\ D_1\ D_2\ D_3\ D_4\ 0\ 0\ 0\ 0\ 0\ 0\ 0]^T$$

$$\mathbf{S_3} = [0\ 0\ 0\ 0\ 0\ 0\ 0\ d_1\ d_2\ d_3\ d_4\ d_5\ d_6\ d_7\ d_8]^T - - - (21)$$

In the case of sixteen input data and fourth Daubechies, multiresolution indicates from Level 0 to Level 3.

The two dimensional wavelets spectrum S is obtained from

$$S = W_n \cdot X \cdot W_m^T \qquad ---(22)$$

Where, W_m^T is a transpose matrix of W_n . From Eq.(22), the discrete inverse wavelets transform is expressed by

$$\mathbf{X} = \mathbf{W_n}^{\mathsf{T}} \cdot \mathbf{S} \cdot \mathbf{W_m} \qquad ----(23)$$

In general, in the case that input data is 2^n and Daubechies function is kth, the algorithm to obtain the frequency levels is shown in Fig.1. The final wavelets spectrum is obtained after the wavelets transform in Eq.(14) continues until the number of final summation elements is less than k.

In this study, 16th Daubechies function is used instead of the fourth Daubechies function that is explained the above as analyzing wavelets. Scaling coefficients of the first line in Eq.(6) are shown in Fig.2. The fundamental properties are similar to fourth Daubechies function. In the case that 16th Daubechies function and the space data number 32 (all space 9.6mm; pseudo-space are included) and the time data number 512 (all measurement time 51.2ms), the wavelets level of the multiresolution is decomposed from level 0 to level 6 in the time, and from level 0 to level 3 in the space.

The absolute values by Fourier transform to the analyzing wavelets of this case are shown in Fig.3 and Fig.4. The each level operates a kind of band pass filter.

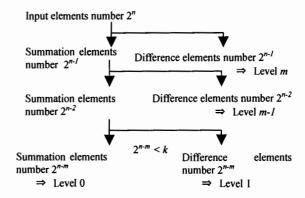


Fig. 1 Algorithm of discrete wavelets transform

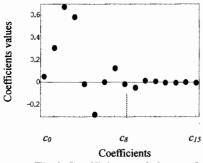


Fig.2 Coefficients of sixteenth Daubechies

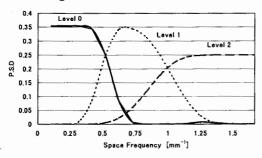


Fig.3 Space frequency of sixteenth Daubechies

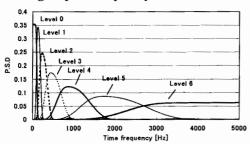


Fig.4 Time frequency of sixteenth Daubechies

EXPERIMENTS

Micro Shear Stress Imaging Chip

A micro shear-stress imaging chip, which is composed of multiple thermal type sensors [Jiang et al. 1996], is shown in Fig.5. The imaging chip has three rows of micro sensors, which ontain an array of 25 sensors. Fig. 6 shows a top and a cross sections of the micro shear stress sensor. Each micro sensor consists of a 150 µm long, 3 µm wide and 0.45 µm thick polysilicon resistor wire, and a 1.2 μm thick and 200 x 200 μm² silicon nitride diaphragm that seals off a 2 µm deep vacuum cavity underneath. The purposes of the cavity are to reduce heat transfer from the resistor wire to the substrate and to increase the sensitivity of the sensor [Huang et al. 1996]. The sensors are connected to external constant temperature mode circuits which are used in a hot-wire anemometer, to drive at a 1.1 overheat ratio, through gold bonding wires. Output from the anemometer circuits is digitized by a 64-channel Keithly Metrabyte ADC board in a Pentium based PC. The sensitivity of the shear stress sensor is about 1.0 V/Pa with a frequency response of 25 kHz at a 10 gain.

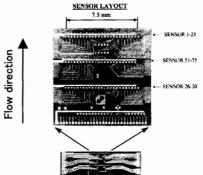


Fig.5 A surface shear stress imaging chip

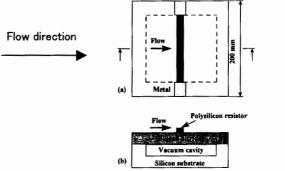


Fig.6 Schematic top (a) and cross-sectional (b) views of the micro shear stress sensor

Experimental Setup

This experiment was carried out in a turbulent channel flow facility. The channel, constructed of 13 mm Plexiglas, is 610 mm x 25.4 mm in a cross-section and 4880 mm long. An axial blower controlled by a DC power generates the air flow in the channel. Previous hot-wire measurement indicates that the channel flow, at a centerline velocity of 10 m/s, consists of a laminar entrance flow and a fully developed turbulent flow at the downstream half channel. The micro shear stress imaging chip was flush mounted into the wall at 4267 mm from the channel inlet where a fully developed turbulent channel flow exists. One array consisting of 25 micro shear-stress sensors that covers a 7.5 mm distance measures the instantaneous spanwise distribution of the turbulent surface shear stress. The Re number (= hu_{∞}/ν , where h is half width of the channel, u_{∞} is centerline velocity) ranged from 6,960 to 17,400.

Shear-Stress Distribution

Statistics values of a wall shear stress measured by this imaging chip are confirmed [Kimura 1999] to be similar to the data obtained by the previous experiments [Obi 1995] [Alfredson 1998], and those by the numerical computation [Kim 1987]. Fig.7 shows the contour of 2-D shear stress distributions measured by the imaging chip. The horizontal axis covers a 7.5 mm wide area (data number 25), and the vertical axis dose

512.2 ms area in time (data number 512) in three different *Re* numbers. Each shear stress is normalized by

$$\frac{\overline{\tau}}{\tau} = \frac{\tau - \tau_{\min}}{\tau_{\max} - \tau_{\min}}$$

using the maximum value τ_{max} and minimum value τ_{min} in each Reynolds number as shown in these figures (a), (b) and (c). In other words, the shear stress values are normalized with the minimum value 0.0 and the maximum value 1.0 in each Re number. The areas with high shear stress are marked by lightgray while dark-gray represents low shear stress with 11 contour Fig.7(d) indicates the original data before the normalization in vertical direction at the point where maximum τ max appears. From Fig.7(d), the transverse scale of the longitudinal high shear-stress streaks varies with Re numbers. The streaks are narrower and packed more densely as Re number increases. They also appear at the shorter time interval as Re number increase. The shear stress fluctuation near wall increases rapidly, resulting in a steep peak in a turbulence boundary layer, because a velocity gradient is rapidly increases e to the influence of eddy guided by bursting events. From rig.7(a) to (c), an area of a white part (a relatively large value) in Re=17400 is smaller and an area of a black part (a comparatively small value) is larger than the lower Re number. The peak of shear stress is smoothed in Re number because the areas are normalized with the maximum and minimum values.

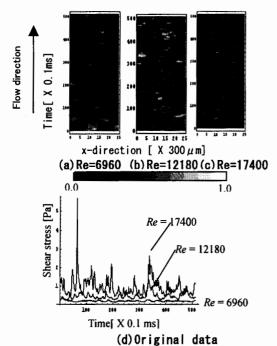


Fig.7 Shear stress measured with the imaging chip

ANALYSIS AND DISCUSSION Analysis Method

The multiresolution analysis is carried out to the wavelets spectrum by the inverse wavelets transform in Eq.(23), after the wavelets transform in Eq.(22) is operated to 25x512 shear stress

distribution data in Fig.7. Zero is put into the matrix position from 26 to 32 in x coordinate in Fig.7 intentionally because the data number of discrete wavelets transform is subject to nth power of two. In the case of sixteenth Daubechies function and 512 (=2 9) input data, the multiresolution classifies to seven levels as shown in

$$\begin{split} &X = [W_{n}^{(5)}]^{T} \cdot S \cdot [W_{m}^{(5)}] = [W_{n}^{(5)}]^{T} \cdot S_{0} \cdot [W_{m}^{(5)}] \\ &+ [W_{n}^{(5)}]^{T} \cdot S_{1} \cdot [W_{m}^{(5)}] + [W_{n}^{(5)}]^{T} \cdot S_{2} \cdot [W_{m}^{(5)}] \\ &+ [W_{n}^{(5)}]^{T} \cdot S_{3} \cdot [W_{m}^{(5)}] + [W_{n}^{(5)}]^{T} \cdot S_{4} \cdot [W_{m}^{(5)}] \\ &+ [W_{n}^{(5)}]^{T} \cdot S_{5} \cdot [W_{m}^{(5)}] + [W_{n}^{(5)}]^{T} \cdot S_{6} \cdot [W_{m}^{(5)}] & --(25) \ . \end{split}$$

 $W^{(5)}$ indicates the five times operation to obtain Daubechies matrix from a matrix C in Eq.(6). The values from 1 to 25 in x coordinate are extracted after the calculation.

Analysis Results

Figs. 8, 9 and 10 (see the final page) show the multiresolution with 11 contours in the low Re number, the middle Re number and the high Re number, respectively. The low shear stress is shown with black, the high shear stress is shown with white from -0.2 to 0.4. The minus values are produced because they are normalized beforehand with the minimum value 0.0. Level 6 is not shown because the noise is dominant. The patterns adding all levels from Level 0 to Level 6 recover completely the original shear stress distributions in Fig.7, because Daubechies analyzing wavelets are orthonormal functions. The relation between the representative frequency, which is calculated with the maximum P.S.D. in Figs.3 and 4, and each wavelets level is shown in Table 1. From these figures, the original input data are decomposed from a low frequency component Level 0 to a high frequency component Level 5. The stripe structures due to a series of bursting events in the low speed shear stress area are visualized clearly on each frequency level without erasing the time and space information. In particular, the light and shade appear more clearly on Levels 0 and 1 in low Re number than high Re number. The shear stress distribution appears until Level 3; however, the wave pattern shades on Levels 4 and 5.

Table 1 Relation between wavelets level and representative frequency

| | Time frequency | Space frequency |
|--|--------------------------------|---|
| Representative frequency on Level 0 | 2.0 ×10¹[Hz] | 1.04 ×10 ⁻¹ [mm ⁻¹] |
| Representative frequency on Level 1 | l.4 ×10²[Hz] | 7.00 ×10 ⁻¹ [mm ⁻¹] |
| Representative frequency on Level 2 | 2. 3 ×10 ² [Hz] | |
| Representative frequency on Level 3 | 4.5 ×10 ² [Hz] | |
| Representative frequency on Level 4 | 8. 8 × 10 ² [Hz] | 1. 2 ×10°[mm ⁻¹] |
| Representative frequency on Level 5 | 1. 7 ×10³[Hz] | |
| Representative frequency on Level 6 | 4.7 ×10 ³ [Hz] | |
| Maximum frequency on Level 6 fmax | 4.7 ×10 ³ [Hz] | 1.6 ×10°[mm ⁻¹] |

Discussion By Kolmogorov Turbulence Theory

The relation between wavelets level and wavenumber is discussed by applying Kolmogorov turbulence theory [Champman 1979] to the time axis on discrete wavelets multiresolved resolution. Kolmogorov wavenumber k_k that indicates the highest viscous dispersion rate is expressed by

$$k_k = \left(\frac{\varepsilon}{v^3}\right)^{\frac{1}{4}} \quad ----(26).$$

Where, ν is a kinematic viscosity, ϵ is an energy transportation rate, that is,

$$\varepsilon = A \frac{v_0^3}{l_0}$$
 $v_0 = 0.2 v$ (27).

Where, v_0 is a representative velocity fluctuation that is assumed to be 0.2 times of the mean velocity v, and l_0 is a length scale in the energy contain region that is assumed to be a radius of the pipeline (=D/2). In this study, the constant A is assumed to be 1.0. The representative wavenumber on wavelets level m is obtained by

$$k_m = \frac{2\pi f_R}{v_0} \tag{28}$$

Where, f_R is a representative frequency which is the maximum power spectrum density in Fig.4. The relation between the wavelets level and the representative wavenumber is shown in Table 2. From this table, Level 0 indicates a wavenumber under the energy contain region, and Level 6 indicates the vicinity of Kolmogorov wavenumber k_k . The wavelets level covers from the energy contain region to the viscous dispersion region. In next, all shear stress on each wavelets level are defined by

$$\overline{\tau_{all}} = \sum_{j=1}^{512} \sum_{i=1}^{25} \sqrt{(\overline{\tau_{ij}})^2}$$
 (29)

to consider the relation between the stress and Re number. i is a position on the time axis, j is a position on the space axis. Fig.11 shows the results. The horizontal axis is the representative wavenumber in Table 1 normalized with Kolmogorov wavenumber k_k .

In this figure, the shear stress gets small in the energy contain region equivalent to Level 0 as *Re* number increases. From 'ig.7(d), the slope of shear stress is steep at high *Re* number in the energy contain region; therefore, the value becomes small when it is normalized. On the other hand, the shear stress gets large in the viscous dispersion region equivalent to Levels 4 and 5 as *Re* number increases. That is because the eddy dispersion is less influenced by the viscosity in high *Re* number. This visualization technique is realized to be proper from the abovementioned qualitative consideration.

Table 2 Relation between wavelets level and representative wavenumber

| | v=8m/s Re=6960 | v=14m/s Re=12180 | v=20m/s Re=17400 |
|--------------------------|--------------------------------------|---------------------------------------|---------------------------------------|
| Representative | 7.7 | 4. 4 | 3. 1 |
| wavenumber on Level 0 | $\times 10^{-1} [cm^{-1}]$ | ×10 ⁻¹ [cm ⁻¹] | ×10 ⁻¹ [cm ⁻¹] |
| Wavenumber in energy | 8.0 | 8. 0 | 8, 0 |
| contain region ko | $\times 10^{-1} [cm^{-1}]$ | ×10 ⁻¹ [cm ⁻¹] | ×10 ⁻¹ [cm ⁻¹] |
| Representative | 5. 4 | 3. 1 | 2. 1 |
| wavenumber on Level 1 | ×10°[cm ⁻¹] | ×10°[cm⁻¹] | ×10°[cm-1] |
| Representative | 9. 2 | 5.3 | 3. 7 |
| wavenumber on Level 2 | ×10°[cm-1] | ×10°[cm ⁻¹] | ×10 ⁰ [cm ⁻¹] |
| Representative | 1.8 | 1, 0 | 7. 1 |
| wavenumber on Level 3 | ×10 ¹ [cm ⁻¹] | ×10 ¹ [cm ⁻¹] | ×10 ¹ [cm ⁻¹] |
| Representative | 3, 4 | 2. 0 | 1.4 |
| wavenumber on Level 4 | ×10 ¹ [cm ⁻¹] | ×10 ¹ [cm ⁻¹] | ×10 ¹ [cm ⁻¹] |
| Representative | 6.7 | 3. 9 | 2. 7 |
| wavenumber on Level 5 | ×10 ¹ [cm ⁻¹] | $\times 10^{1} [cm^{-1}]$ | $\times 10^{1} [cm^{-1}]$ |
| Representative | 1. 9 | 1.0 | 0. 7 |
| wavenumber on Level 6 | $\times 10^{2} [cm^{-1}]$ | $\times 10^{2} [cm^{-1}]$ | $\times 10^{2} [cm^{-1}]$ |
| Maximum wavenumber on | 1. 96 | 1. 12 | 0. 79 |
| Level 6 k _{max} | $\times 10^{2} [cm^{-1}]$ | $\times 10^{2} [cm^{-1}]$ | $\times 10^{2} [cm^{-1}]$ |
| Kolmogorov wavenumber | 3. 49 | 2. 67 | 1. 76 |
| k_k | $\times 10^{2} [cm^{-1}]$ | $\times 10^{2} [cm^{-1}]$ | $\times 10^{2} [cm^{-1}]$ |

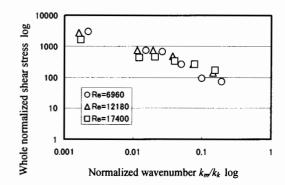


Fig.11 Relation between whole normalized shear stress and wavenumber on the time-space

CONCLUSIONS

A micro machine shear-stress imaging chip was used to measure the instantaneous shear-stress distribution in a turbulent wall boundary layer in conditions from *Re* number 6,980 to 17,400. The two dimensional shear stress distributions are visualized with discrete wavelets transform. As a result, the following becomes clear.

- (1) Structures of wall shear stress are able to be decomposed and be clearly visualized on each frequency level without erasing time-space information.
- (2) The whole shear stress on Levels 0 and 1 relevant to near energy contain region in high *Re* number is smaller than that in low *Re* number. The slope of shear stress is steep at high *Re* number in the energy contain region; therefore, the value becomes small when it is normalized.
- (3) The whole shear stress on Levels 4 and 5 relevant to near viscous dispersion region in high *Re* number is higher than that in low *Re* number. It is reasonable because the structure is less influenced by the viscosity in high number.

ACKNOWLEDGMENTS

The authors wish to thank Dr. F. Jiang and Dr. Y.-C. Tai from the California Institute of Technology.

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