

# Synthesizing Turbulent Channel Flow

## A Proposed Form of Solution to Steady Turbulence in Channels

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**Abstract:** There has long been a need for a fully theoretical basis for describing turbulent flows. We propose a formal representation of the random velocities using a few ordinary smooth non-random functions together with an ordinary Stochastic Integral. We call this the **Gaussian Transform** or **GXF**. For example, for **3D** Channel Flow, we need only two **CDF**'s (Cumulative Distribution Functions) and four ordinary correlation functions. The simpler case of **2D** Channel Flow requires only one **CDF** and one correlation function. Decaying Isotropic Turbulence (**HIT**) requires only one **CDF** and two correlations functions. We will develop these functions for **2D** Channel Flow; we will describe how to compute them numerically from **DNS** results; we will demonstrate the synthesis of multiple realizations of the flow; and we will apply the **NSE** to this representation to come to a full solution to **2D** Channel Flow.

**Keywords:** Theoretical Turbulence, Multidimensional Stochastic Integrals, Ritz-Galerkin approximation, Metalog **CDF** approximations, Stationary Ergodic Stochastic Processes

## 1 Introduction

This paper is divided into two main parts and a conclusion section:

First a theoretical development of the **GXF** method for **2D** Channel Flows with adequate (perhaps heuristic) rigor. We will use the **CDF**; the corresponding quantile functions; and then the corresponding normal random variable. From this we can compute the kernel of the **Wiener Stochastic Convolution Integral**.

Second, we determine numerically the two functions using **DNS** results. The **CDF** is obtained from recorded velocities  $\mathbf{u}[\mathbf{i}, \mathbf{x}, \mathbf{y}, \mathbf{t}]$  (Python notation) by simple algebra or perhaps metalog mechanisms, thence the numerical quantile function, thence the numerical Auto-Correlation, thence the kernel of the Wiener Integral.

In conclusion, we discuss analytic representation of the **CDF** and auto-correlation functions. These can be plugged into the **NSE** and optimized using (say) Ritz-Galerkin methods for a theoretical solution to this **2D** Channel Flow Problem. There are many more flow situations which are amenable to this analysis, including decaying flows, **MFD** flows, compressible flows, and many other internal and external flows.

## 2 Developing the Gaussian Transform

Consider **2D** Channel Flow with the usual assumptions: Newtonian Fluid, no-slip walls, incompressible, obeying the Navier Stokes equations, etc. The flow is stable, stationary on  $\mathbf{x}$  (stream-wise) and  $\mathbf{t}$ . Reynold's averaging applies, i.e., the flow is "**ergodic**" on  $\mathbf{x}$  and  $\mathbf{t}$ .

To illustrate the process: place a velocity probe at some point  $\{\mathbf{x}_0, \mathbf{y}_0\}$  and record the velocity  $\mathbf{u}(\mathbf{t})$  against time. This random  $\mathbf{u}(\mathbf{t}, \alpha)$  (including a sampling parameter  $\alpha$ ) has a smooth **CDF** so that:

$$\begin{aligned} \Pr[u(t, \alpha) \leq \lambda] &= CDF(\lambda) && \text{Smooth Strictly Increasing} \\ q(t, \alpha) &= CDF(u(t, \alpha)) && \text{Derived Quantile RV, has linear cdf} \\ \phi(t, \alpha) &= \Psi^{-1}(q(t, \alpha)) && \text{Derived Normal RV, has Normal cdf} \\ \phi(t, \alpha) &= \int \Phi(t - \tau) dr(\tau, \alpha) && \text{Wiener Stochastic Convolution Integral} \end{aligned} \tag{1}$$

Here  $\Psi^{-1}(\lambda)$  is the inverse Normal distribution. The  $\Phi(\tau)$  kernel is determined uniquely from the auto-correlation of  $\mathbf{u}[\mathbf{t}, \alpha]$ . Thus, given a **CDF** and an auto-correlation function, any stationary ergodic random variable can be synthesized by reversing the process of equation (1) thus:.

$$\begin{aligned}
\phi(t, \alpha) &= \int \Phi(t - \tau) dr(\tau, \alpha) && \text{Wiener Stochastic Convolution Integral} \\
q(t, \alpha) &= \Psi(\phi(t, \alpha)) && \text{Generated Normal-Quantile RV, has linear cdf} \\
u(t, \alpha) &= CDF^{-1}(q(t, \alpha)) && \text{Generated Physical RV, has physical cdf}
\end{aligned} \tag{2}$$

By a straightforward extension, **2D** Channel Flow can be synthesized as follows

$$\begin{aligned}
\phi(x, y, t, \alpha) &= \int \Phi(x - \varepsilon, t - \tau, y) dr(\varepsilon, \tau, \alpha) && \text{2D Wiener Integral, Normal RV} \\
q(x, y, t, \alpha) &= \Psi(\phi(x, y, t, \alpha)) && \text{Generated Normal-Quantile RV} \\
s(x, y, t, \alpha) &= CDF^{-1}(q(x, y, t, \alpha), y) && \text{Generated Physical Function RV} \\
u^x(x, y, t, \alpha) &= -\partial_y s(x, y, t, \alpha) && \text{u is "curl" of s} \\
u^y(x, y, t, \alpha) &= +\partial_x s(x, y, t, \alpha)
\end{aligned} \tag{3}$$

So, we claim, **2D** channel turbulence is completely described by these two functions: **CDF( $\lambda, y$ )** and  **$\Phi(x, t, y)$**

The extension to **3D** Channel Flow turbulence is straightforward but complex in detail. The **3D** case requires two **CDF( $\lambda, y$ )** and four  **$\Phi(x, t, z, y)$** .

### 3 Using the Gaussian Transform Numerically and Analytically

A robust **DNS** run for **2D** Channel Flow turbulence will provide the data for a numerical description of both the **CDF( $\lambda, y$ )** and  **$\Phi(x, t, y)$** . There is much work to be done to compute these two functions over all **y**-values. And then for several **Re** values. But the results will allow synthesis of any realization of the flow.

Also, the **CDF**'s can be approximated by a variety of techniques – most promising is **Metalog** (see Wikipedia) with a few parameters. And the auto-correlation functions can be approximated by a sum of Hermite Functions or other rapidly decreasing functions – hopefully with a few parameters. This will give good insight into the structure of both functions against **y**-values and **Re**-values.

A third prospect is this: construct the approximation just above (or better); insert into the **NSE** and minimize the mean-square of the residual – a **Ritz-Galerkin** model. Hopefully a few parameters will suffice, and the results will be close to the **DNS** results.

Much more to be said. Many results forthcoming.

### 4 Conclusion and Future Work

We have shown that there is an **Analytic Representation of Stationary Turbulent Channel Flows**. We have outlined the construction of this mechanism using **2D** Channel Flow as an exemplar. The methods apply to a variety of internal and external flows.

We will continue the investigation using **2D** Channel Flows to determine dependence on **Re** and to develop better approximations to **CDF( $\lambda, y$ )** and  **$\Phi(x, t, y)$** . Ultimately – time and energy permitting – we will apply the **NSE** to this formulation most likely using a Ritz-Galerkin best fit technique.

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