

1 Mean as a random variable

1.1 Normal distribution

We define a n-sample as n realizations of a normally distributed random variable.

1. Write a **for** loop that you iterate **nsamp** times in which you define a n-sample and compute its mean \bar{x} . Use $n = 500$.
2. Use **nsamp**=4000 and plot the PDF of \bar{x} .
3. \bar{x} is itself a random variable. What is its mean? its standard deviation $\sigma_{\bar{x}}$? Does the distribution look Gaussian?
4. Check how the pdf and standard deviation $\sigma_{\bar{x}}$ vary with n .

1.2 chi-squared distribution

We now redo exactly the same computation using a random variable following a chi-squared distribution

1. Write a function to generate a random variable following a chi-squared distribution with $l = 10$ degrees of freedom

2. Write a **for** loop that you iterate **nsamp** times in which you define a n-sample (following a chi-squared distribution) and compute its mean \bar{x} . Use $n = 500$.
3. Use **nsamp**=4000 and plot the PDF of \bar{x} . Does the distribution of the means still look Gaussian?
4. Check how the pdf and standard deviation $\sigma_{\bar{x}}$ vary with n .

1.3 Cauchy distribution

We now redo exactly the same computation using a random variable following a Cauchy distribution

1. Write a function to generate a random variable following a Cauchy distribution
2. For different n , compute the estimated standard deviation and show that it's diverging with n , the number of samples.
3. Write a **for** loop that you iterate **nsamp** times in which you define a n-sample (following a Cauchy distribution) and compute its mean \bar{x} . Use $n = 500$.
4. Use **nsamp**=4000 and plot the PDF of \bar{x} . Does the distribution of the means still look Gaussian?
5. Check how the pdf and standard deviation $\sigma_{\bar{x}}$ vary with n .

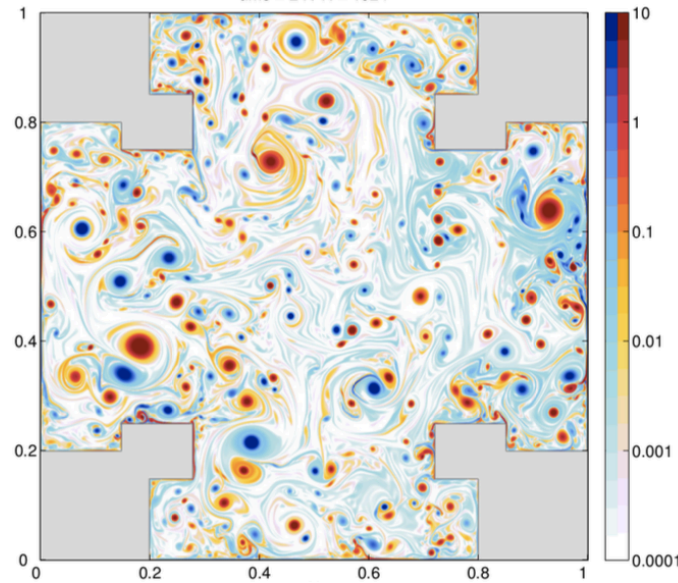
2 Standard deviation as a random variable

We define a n-sample as n realizations of a normally distributed random variable as in the first part, and redo the same computation but for σ the estimated standard deviation of x .

1. generate **nsamp**=4000 values of the of random variables **x=randn(n,m)** ;

2. compute the m standard deviations : `sigma=std(x,0,1);`
3. plot the pdf of `sigma`, its cdf and print its first four moments.
4. Check how the pdf varies with n .

3 Intermittency



Not all random variables are gaussian. Here is an example of a very intermittent variable. Load the `turbulence2D_with_boundaries.mat` file. It contains 4 timeseries K (the total kinetic energy), V (the total enstrophy) and their time derivatives with

$$K = \frac{1}{2} \int_{\mathcal{D}} (u^2 + v^2) dA , \quad (1)$$

$$V = \frac{1}{2} \int_{\mathcal{D}} \zeta^2 dA \quad (2)$$

and ζ the vorticity. These timeseries come from a numerical simulation of decaying 2D turbulence in a closed domain \mathcal{D} with friction on the boundaries.

1. plot K and dK/dt as a function of time [use distinct plots]. Are the series stationary?
2. same for V and dV/dt .
3. We will focus on the time derivatives. Define x as dV/dt over the time interval $[2.10^4 \ 3.10^4]$. [use `find`]
4. Plot the histogram of x . Redo the same plot in semilogy axis.
5. Compute the mean, standard deviation, skewness and the kurtosis of x .
6. Superimpose the gaussian that has the same mean and the same standard deviation. Compare the tails (extreme fluctuations).