

Lesson 3: ESTIMATION OF THE TURBULENT DIFFUSION COEFFICIENTS

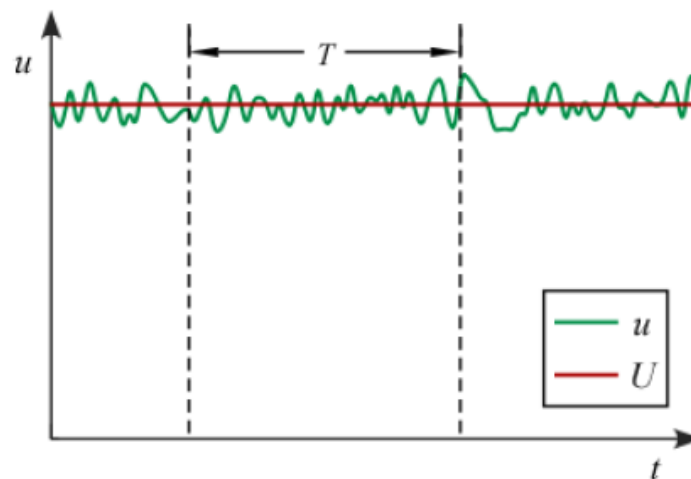
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In presence of turbulent velocity field, the velocity $\mathbf{u}$ can be expressed as proposed by Reynolds:

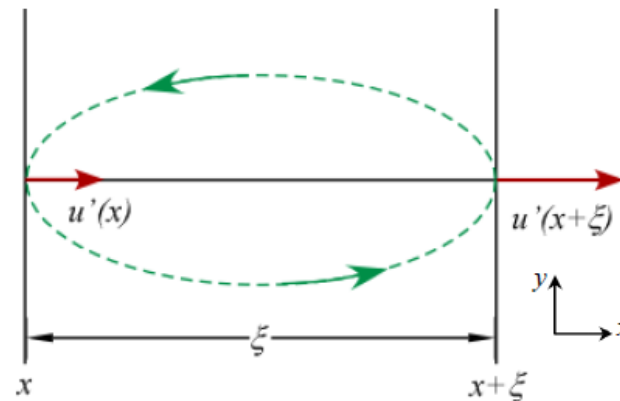
$$\mathbf{u} = \mathbf{U} + \mathbf{u}'$$



$$\mathbf{U} = \langle \mathbf{u} \rangle \text{ mean velocity} \Rightarrow U = \frac{1}{T} \int_0^T u \, dt \quad \text{mean}(u) ; \text{ or } \text{trapz}(t, u) / T ;$$

$$\mathbf{u}' \text{ velocity fluctuations} \Rightarrow \mathbf{u}' = \mathbf{u} - \langle \mathbf{u} \rangle$$

The vortex affects the magnitude of the velocity fluctuation



To understand the time evolution of the vortices, we define the *Eulerian Temporal Correlation Coefficient* as follows:

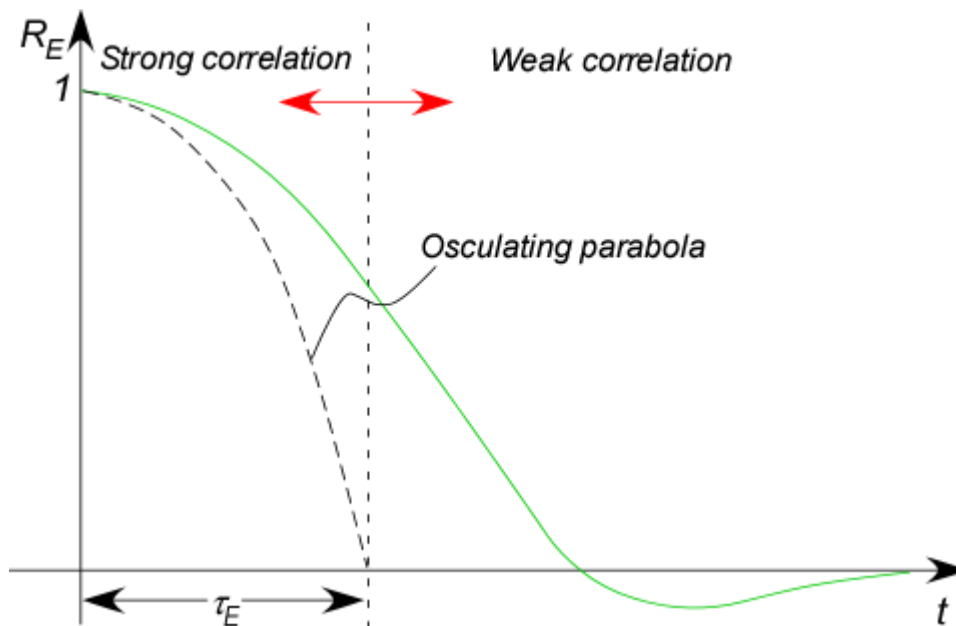
$$R_E(x, t, \tau) = \frac{\langle u'(x, t) u'(x, t + \tau) \rangle}{\sqrt{\langle u'^2(x, t) \rangle} \sqrt{\langle u'^2(x, t + \tau) \rangle}}$$

If the process is *ergodic*:

$$\Rightarrow R_E(x, \tau) = \frac{\langle u'(x) u'(x, \tau) \rangle}{\sqrt{\langle u'^2(x) \rangle} \sqrt{\langle u'^2(x, \tau) \rangle}}$$

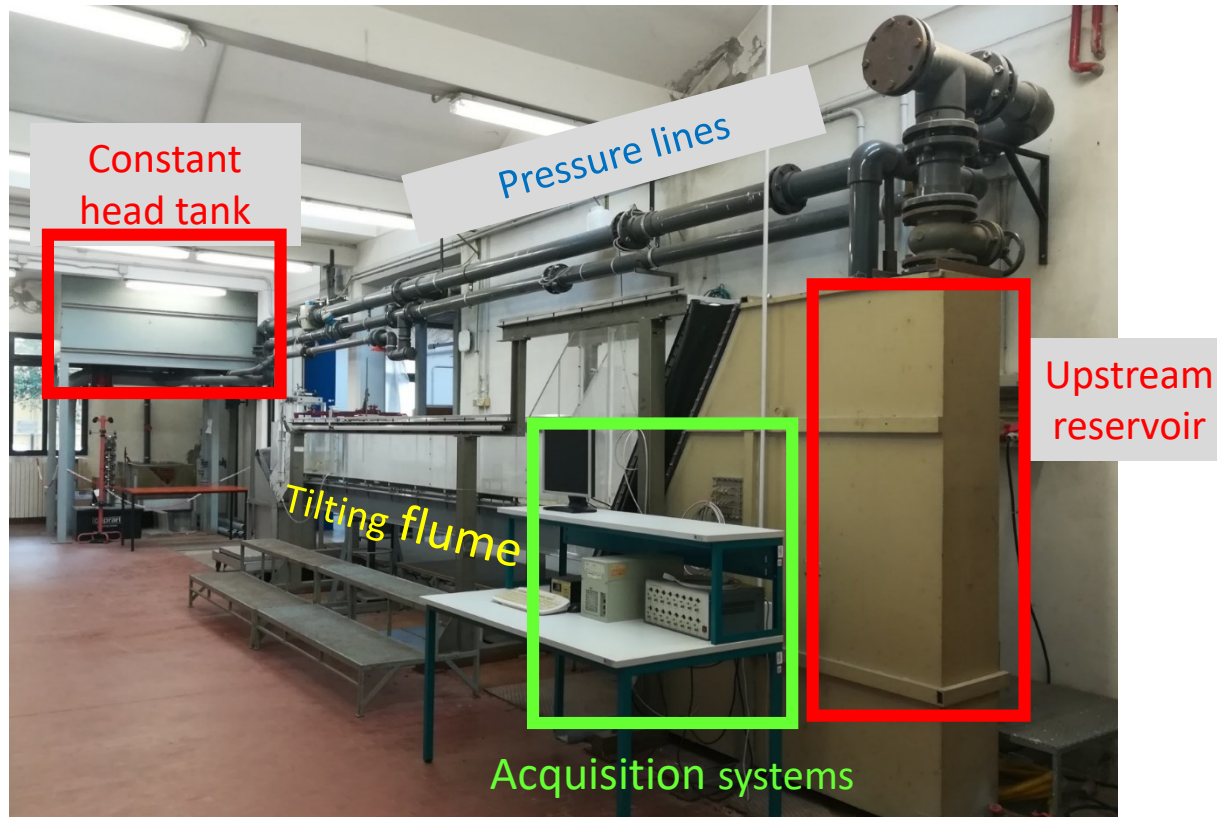
The correlation function has the following properties:

$$\begin{cases} R_E(0) = 1 \\ R_E(t) = R_E(-t) \\ R_E(t) \leq 1 \end{cases}$$

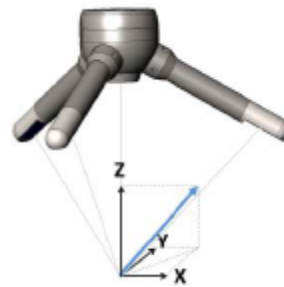


$T_{Ei} = \int_0^\infty R_E(\tau) d\tau$ is the time scale of macrovortex in a selected direction i

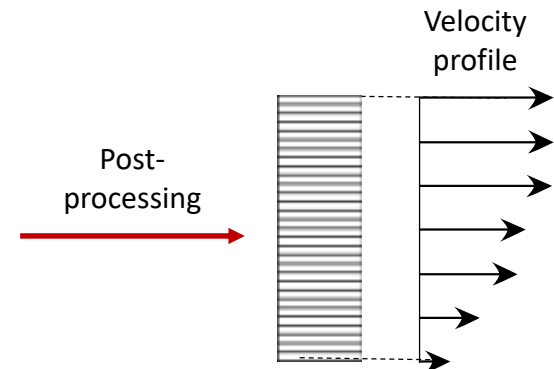
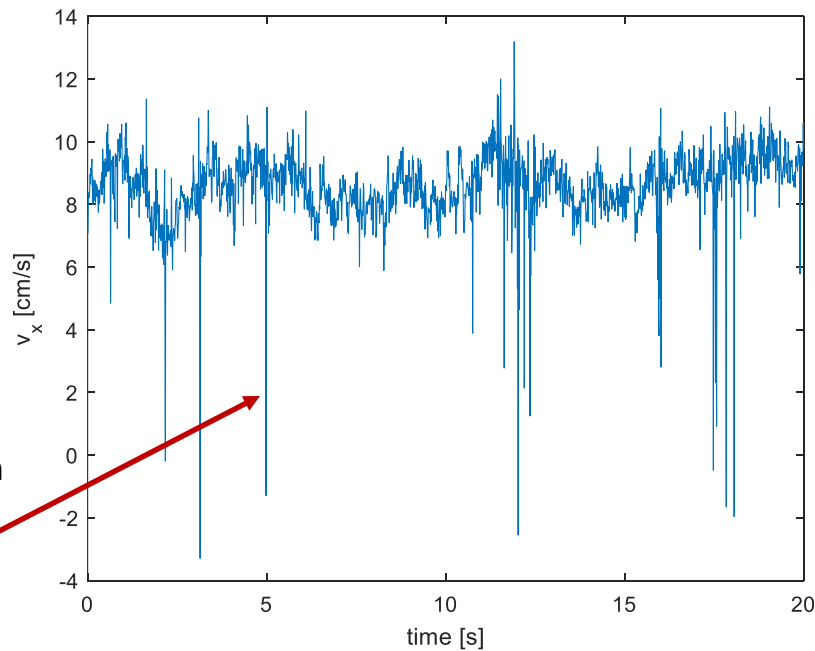
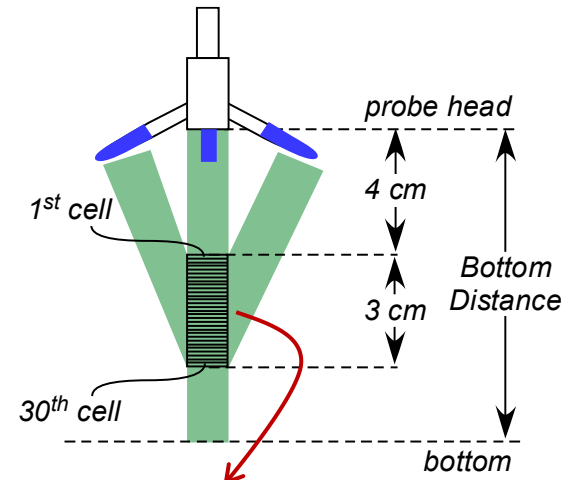
$e_i = \langle u_i'^2 \rangle T_{Ei}$ is the diffusion coefficient for $t \gg T_{Ei}$



Q [L/s]	Water depth [cm]	Slope [%]
3.0	12	0.1
5.0	13	0.1



The blue arrow indicates a positive velocity



Goal #1:

1. Compute the mean velocities, the fluctuations, and the standard deviation in the 3 directions
2. Compute the mean velocity profile along the vertical direction

Goal #2:

3. Compute the coefficient of temporal autocorrelation ($R_E(x, \tau)$) in at least one cell
4. Compute the time scale of the macrovortex (T_{E_i})

Goal #3:

5. Compute the diffusion coefficient (e_i)
6. Compare e_i with the value provided by the literature, e.g., Reynolds analogy theory

1. Load the velocity data
2. Extract the velocities (`Vx = Data.Profiles_VelX`, `Vy = Data.Profiles_VelY`, `Vz = Data.Profiles_VelZ`) (in the file the velocities are in [m/s], attention to the units!)
3. Extract the distance between the bottom and the head of the sensor (`Data.BottomCheck_BottomDistance`) (consider the mean value, in the file it is in [m])
4. Define the vector of cells' position to be used to plot the mean velocity (remember the definition of the ADV and how it measures the distances: 30 cells in 3 cm of sampling volume)

For each cell:

5. Compute the 3 mean velocity components (`mean()`)
6. Compute the fluctuations for each direction (remember the formula!)
7. Compute the standard deviation for the 3 velocities ($SD = \sqrt{\langle u_i'^2 \rangle} = \sqrt{\frac{\sum_{i=1}^N u_i' \cdot u_i'}{N}}$, hint: use the matrix operation or for cycle)

For the longitudinal direction:

8. Plot the mean velocity

For 1 or more cells:

1. Load the velocity data (v_x, v_y, v_z)
2. Define the time of measurements: $f = 100 \text{ Hz}$, $T_{end} = \Delta t \cdot N$
3. Create the time vector (`linspace(...)`)
4. Divide the time vector in two: the first half is used for the analysis (we will see later why we need to divide the data in two vectors!)

For each velocity vector:

5. Compute the mean velocity (`mean()` or `trapz(t, v) / T`)
6. Compute the fluctuations (or extract from the vectors computed for *goal #1*)
7. Compute and plot $R_E(x, \tau) = \frac{\langle u'(x)u'(x, \tau) \rangle}{\sqrt{\langle u'^2(t) \rangle} \sqrt{\langle u'^2(x, \tau) \rangle}}$
8. Compute $T_E = \int_0^\infty R_E(\tau) d\tau$

7. Compute $R_E(x, t, \tau) = \frac{\langle u'(x, t) u'(x, t + \tau) \rangle}{\sqrt{\langle u'^2(x, t) \rangle} \sqrt{\langle u'^2(x, t + \tau) \rangle}}$

$\langle u'(x) u'(x, \tau) \rangle$

Ergodic process (remember that usually x is fixed!)

$u'_{x, y, z}(i)$

where i defines the time

$u'_{x, y, z}(i + j)$

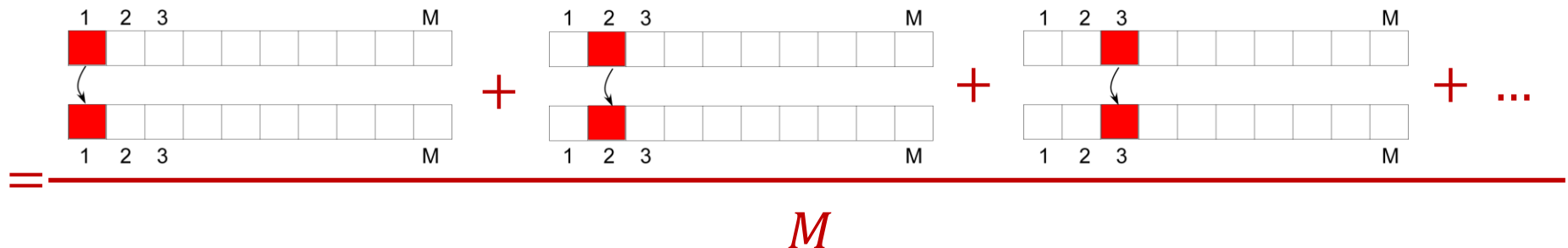
where j resembles $\tau = 0, \tau = \Delta t, \tau = 2\Delta t, \dots, \tau = M\Delta t$

Remember that you're using half of the data!

$\langle u'(x) u'(x, \tau) \rangle \Rightarrow \langle \dots \rangle$ remember that it means the time averaged value!

Where:

$\tau = 0 \Rightarrow \langle u'(x) u'(x) \rangle =$



7. Compute $R_E(x, t, \tau) = \frac{\langle u'(x, t) u'(x, t + \tau) \rangle}{\sqrt{\langle u'^2(x, t) \rangle} \sqrt{\langle u'^2(x, t + \tau) \rangle}}$

$\langle u'(x) u'(x, \tau) \rangle$

Ergodic process (remember that usually $\mathbf{x}$ is fixed!)

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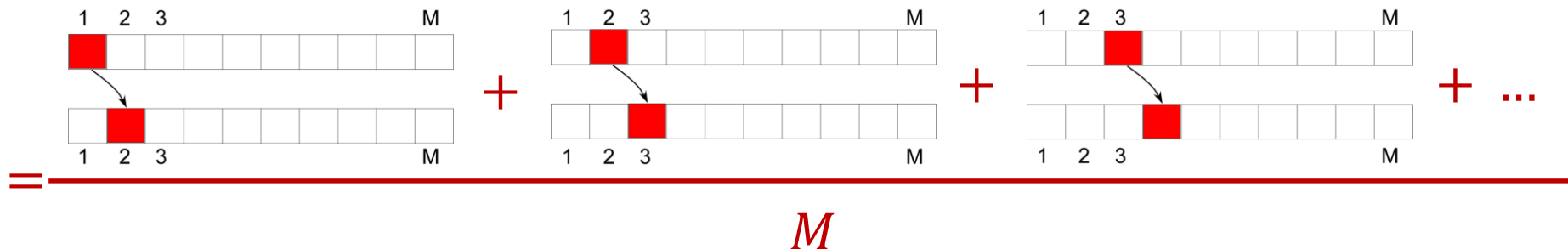
where j resembles $\tau = 0, \tau = \Delta t, \tau = 2\Delta t, \dots, \tau = M\Delta t$

Remember that you're using half of the data!

$\langle u'(x) u'(x, \tau) \rangle \Rightarrow \langle \dots \rangle$ remember that it means the time averaged value!

Where:

$\tau = \Delta t \Rightarrow \langle u'(x) u'(x + \Delta t) \rangle =$



7. Compute $R_E(x, t, \tau) = \frac{\langle u'(x, t) u'(x, t + \tau) \rangle}{\sqrt{\langle u'^2(x, t) \rangle} \sqrt{\langle u'^2(x, t + \tau) \rangle}}$

$\langle u'(x) u'(x, \tau) \rangle$

Ergodic process (remember that usually x is fixed!)

$u'_{x, y, z}(i)$

where i defines the time

$u'_{x, y, z}(i+j)$

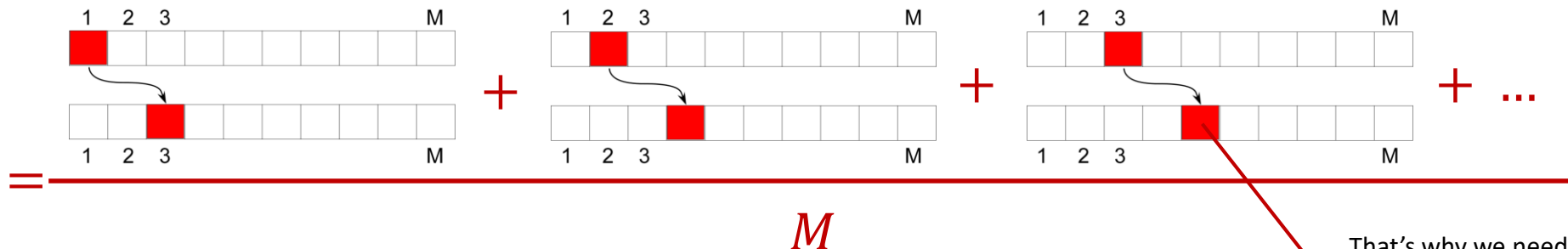
where j resembles $\tau = 0, \tau = \Delta t, \tau = 2\Delta t, \dots, \tau = M\Delta t$

Remember that you're using half of the data!

$\langle u'(x) u'(x, \tau) \rangle \Rightarrow \langle \dots \rangle$ remember that it means the time averaged value!

Where:

$\tau = 2\Delta t \Rightarrow \langle u'(x) u'(x + 2\Delta t) \rangle =$



That's why we need to divide the data in two vectors!

7. Compute $R_E(x, t, \tau) = \frac{\langle u'(x, t) u'(x, t + \tau) \rangle}{\sqrt{\langle u'^2(x, t) \rangle} \sqrt{\langle u'^2(x, t + \tau) \rangle}}$

$\langle u'(x) u'(x, \tau) \rangle$ *Ergodic process* (remember that usually **x** is fixed!)

$u'_{x,y,z}(i)$ $u'_{x,y,z}(i+j)$

where i defines the time where j resembles $\tau = 0, \tau = \Delta t, \tau = 2\Delta t, \dots, \tau = M\Delta t$

Remember that you're using half of the data!

$\langle u'(x) u'(x, \tau) \rangle \Rightarrow \langle \dots \rangle$ remember that it means the time averaged value!

Thus, the numerator is computed as:

$$\Rightarrow \langle u'(i) u'(i + j) \rangle = \frac{1}{M} \sum_j \sum_i u'(i) \cdot u'(i + j)$$

For the denominator, the same methodology has to be applied! (Pay attention that there is the square root and the square of **u'** in the formula!)

7. Compute $R_E(x, t, \tau) = \frac{\langle u'(x, t) u'(x, t + \tau) \rangle}{\sqrt{\langle u'^2(x, t) \rangle} \sqrt{\langle u'^2(x, t + \tau) \rangle}}$

Hint:

1. $\langle u'(i) u'(i + j) \rangle = \frac{1}{M} \sum_j \sum_i u'(i) \cdot u'(i + j)$: use two for loops

```
For j= ...
    for i= ...
```

```
end
```

```
End
```

2. $\langle u'(x, t) u'(x, t + \tau) \rangle$: create three temporary variables (1, 2, and 3)

$\sqrt{\langle u'^2(x, t) \rangle}$ $\sqrt{\langle u'^2(x, t + \tau) \rangle}$

8. Compute $T_E = \int_0^\infty R_E(\tau) d\tau$

Hint:

1. Once you computed R_E , T_E is simply the time integral (pay attention to use the correct time vector! Where is define R_E ?)

For 1 or more cells:

1. You already know how to compute $\langle u_i'^2 \rangle$
2. $e_i = \langle u_i'^2 \rangle T_{E_i}$ (T_{E_i} computed in goal #2)
3. Compare e_i with the value provided by the literature, e.g., Reynolds analogy theory