

2. [housing market, one variable](#)
3. [This](#) is a report [?] in which I use ML for improving wall-functions for Large Eddy Simulations.

Next, look at the Python script *An example of Python code for Machine Learning for optimizing a k-omega turbulence model.* at the [homepage](#). This script uses DNS data to optimize the coefficient C_μ in the $k - \omega$ model where the turbulent viscosity is computed as $\nu_t = C_\mu k / \omega$, $C_\mu = 1$, see Eq. 16.11.

Assignment 1.9. Watch the recorded lecture on Appendix 1, Part II (you can download it from the course [www](#) page, Assignment 1). Try to understand the Python script. I use $\partial \bar{v}_1 / \partial x_2$ as input (influence variable) and C_μ as output (target). When you execute the script, you find that you get different result (i.e. different RMS error) every time: why?

The RMS error is rather large (close to 20%). 80% of the DNS data (randomly chosen) are used for training. Maybe that's a bad idea; maybe it would make sense to only use data for which $|\overline{v'_1 v'_2}|$ is large. Try that. What happens if you now change the C value in the `svr`-call? Does the ε value have any influence?

Assignment 1.10. Now you have developed a new ML $k - \omega$ model. Let's find out if it gives better results! I.e. implement the ML model in a Python CFD code. First you need to export your ML `svr` $k - \omega$ model by saving it to disk and then loading it in your Python CFD code. I show how to do that at pp. 13-14 in [?].

Now, which Python CFD code should you use? I give you three options.

1. In the CFD course in Study Period 2 you wrote a CFD code for fully-developed channel flow. Use that one.
2. You can use the Python code `rans-k-omega.py` which solves the fully-developed flow in a 1D channel (it solves \bar{v} , k and ω).
3. You can use my Python code **pyCALC-RANS** [?]. It can be downloaded [here](#). I recommend that you compute fully-developed channel flow at $Re_\tau = 5200$, see Section 8 in [?]. The turbulent viscosity is computed in module `vist_kom`. At the end of this module, the module `modify_vis` is called which resides in file `modify_case.py`. Here you can re-compute ν_t using your ML $k - \omega$ model.

Assignment 1.11. Here you make a new ML model (and CFD predictions) as I describe in the recorded lecture. You will use $|\overline{v'_1 v'_2}|$ as output and $\partial \bar{v}_1 / \partial x_2$ and ν_t as input.

Assignment 1.12. Finally you make a ML model for the mixing length model as I describe in the recorded lecture.