# **Assignment 4**

Allowed library imports: numpy, matplotlib

# E12. (10 points) Array calculations

Do not use loops over the time dimension for this exercise!

- Create a numpy array called  ${\tt t}$  that represents N=5000 time values between 0 and 60 seconds (both included).
- Create another single numpy array called data, which represents  $\sin(\omega t)$ ,  $\cos(\omega t)$  for the following choices of angular frequency:  $\omega = 2\pi/60$ s,  $2\pi/30$ s,  $2\pi/100$ s,  $2\pi/120$ s.
- Make one nice plot showing all of the lines (both functions, all frequencies) with a descriptive legend.
- Now add two more functions to the data:  $\sin^2$  and  $\cos^2$ , by extending the corresponding dimension in the array. Repeat the plotting, but now showing *only* the newly added quadratic functions.
- Sum all  $\sin$  and  $\sin^2$  functions, for all frequencies, into one new numpy array. Similarly sum all  $\cos$  and  $\cos^2$  results into another array. Plot both, and also their sum, into one plot.

## E13. (10 points) Data creation

Do not use loops for this exercise!

- Create a numpy array that represents dimensions (hour, minute, index, data), for a single day of measurements (hours: 0-23, minutes:0-59), where the index indicates 5 different weather stations and the data 3 data entries per location, filled with zeros of type np.float64. Print the shape and the dtype of your array. How many elements are contained in the array?
- Assume the 3 data entries represent solar irradiance in W/m<sup>2</sup>, temperature in Kelvin and wind speed in m/s at 10 m height, respectively. Set all solar irradiance values to 250 W/m<sup>2</sup>, then replace all values for the time between 9 PM and 5 AM to zero. Check that this worked as intended, by printing appropriate slices.
- Create another numpy array representing dimensions (minute, index) with random values between 3 and 8 m/s. Set all wind speed data for all stations to that data, such that it repeats every hour. Check that this worked as intended, by printing appropriate slices.
- Fill the temperature data with random values between 5 and 20 degrees Celsius (translate that into Kelvin). Station 2 has some problem during nights, so replace all temperatures by np.nan between 9 PM and 5 AM. Station 0 and 1 do not work correctly at high wind speeds, so replace the temperature by np.nan whenever the wind speed is above or equal 6 m/s for those stations. Again, check that this worked as intended.

## E14. (10 points) Wind farm data - part I

The provided data file results\_farm.csv.gz has a single header line, stating the column names:

State, turbine, YAW, WD, AMB\_WS, WS, AMB\_P, P, GY, NY, P75, P90, EFF, WLOSS

The State column contains integer numbers, indexing 30-minute time average data (the time stamp itself has been removed, since it is not needed here).

- Read the file results\_farm.csv.gz into a numpy.ndarray, ignoring the first row (i.e., the above header line). Print the first 10 rows of data. Note that you do not need to extract the gz file before reading it with numpy.
- Select the subset of the data for which the produced power is zero. What is the maximal ambient wind speed AMB\_WS in this subset? What the minimal wake-corrected wind speed WS in another subset with power greater than zero? The data unit in both cases is m/s.

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• How many turbines are in the wind farm? Create a separate numpy array for each turbine (use at most a single loop). Then, find out how much mean power *P* was produced by each turbine with respect to the states. The unit of P in the data is kW. How high is the annual energy production (AEP) in GWh based on this data for the whole farm?

#### E15. (10 points) Wind farm data – part II

- Read the data from E14 into a single numpy array, as described there.
- Create a simple line plot, showing the mean ambient power AMB\_P and the mean produced power P wrt. the states, as a function of turbine index. Include axes labels, a legend and a title.
- Replace all values of the wind turbine efficiency EFF by np.nan if they are zero (or below). Then create a 2D pcolormesh plot that shows the variable EFF as a function of turbine on x-axis and state on the y-axis, with coordinate (0, 0) in the lower left corner. Make the plot as pretty as you can, e.g., add axes labels, a title, and a color bar. Follow the lecture on matplotlib, but please, no red lines between the data squares.
- Similarly, create two beautiful pcolormesh plots showing the wind speed WS and the produced power P. Use different color maps for each of the latter three 2D-plots, and always show color bars. You may also use other ways of visualizing this 2D data than pcolormesh, if you prefer (but please not just a bunch of single lines).

#### E16. (10 points) Topography data

The file RoedeserBerg\_large.pts.gz contains (x,y,z) topography data of a location near Kassel, Germany.

Solve the whole exercise without using loops.

- · Read the data contained in the above file into memory.
- Select the subset of points whose horizontal distance from the position  $p_0=(513196.5,5689695.0)$  is less than radius r=5 km.
- Store the selected data into a new file RoedeserBerg.pts.gz in the same format as the original file. How many points are stored in the file?
- Visualize the selected topography data using matplotlib or a sub-package. Additionally store the resulting graphic into a png file RoedeserBerg.png