Part 1. Objectives

- Interpret behavior of functions by their Fourier components
- Understand how to use interpolants to numerically take derivatives of functions
- Undersand what limits the accuracy of numerical differentiation
- Understand how to derive quadrature rules

Part 2. Splines: Counting conditions

Suppose we would like to construct a piecewise polynomial interpolant consisting of two pieces, one with a quadratic polynomial, and one with a cubic.

How many conditions do we need in addition to the $f(x_i) = y_i$ interpolation conditions?

Part 3. Differentiation: Error prediction

Suppose we numerically take a derivative of a function f(x) using monomials at four equally-spaced points on an interval of length h. The obtained interpolation error is found to be 0.1. What interpolation error do you predict for the same function on a subinterval of the original one that has length h/2?

Part 4. Numerical Differentiation

You are given nodes and data for an interpolant. Evaluate and plot its derivative. INPUT:

- \mathbf{x} , a 4-vector of x coordinates
- f_x , a 4-vector of function values f(x)

OUTPUT:

- fp_x, a 4-vector of point values of the *derivative* of the polynomial interpolant for the data f_x at the nodes x
- coeffs, the coefficients (α_i) for the derivative expressed as a linear combination of monomials: $f'(x) = \alpha_0 \cdot 1 + alpha_1 \cdot x + \alpha_2 \cdot x^2$

Print coeffs[-1] (the last entry of it). What should it be?

```
import numpy as np
import numpy.linalg as la
import matplotlib.pyplot as pt
fp_x =
coeffs =
# plotting code below, no need to modify
fp_interpolant = 0
for i, coeff in enumerate(coeffs):
    fp_interpolant += coeff*xp**i
pt.plot(x, f_x, "o", label="f")
pt.plot(x, fp_x, "o", label="f")
pt.plot(x, fp_interpolant)
pt.legend(loc="best")
```