Problem 1: Consider the data set:

Determine an interpolant p(x) such that $p(x_0) = y_0, p(x_1) = y_1, \dots, p(x_7) = y_7$.

(1) (5 Points) Find a polynomial interpolant as

$$\Pi_n(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0 \qquad n \le 6.$$

Use **Lagrange polynomial** interpolant. Show your results in a plot and plot the fitted polynomial functions together with data points.

- (2) (5 Points) Use linear piecewise interpolation. Show your results in a plot and plot the fitted polynomial functions together with data points.
- (3) (5 points) Use least squares to fit a polynomial function $\Pi_n(x)$. Try both n = 6 and n = 3. Show your results in a plot and plot the fitted polynomial functions together with data points.

Problem 2: Approximate the function

$$f(x) = \frac{1}{1 + 36x^2}$$

in the interval [-1,1] through interpolation.

(1) Find a polynomial interpolant as

$$p(x) = \Pi_n(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$$
.

You can use Lagrange polynomial to construct the interpolant.

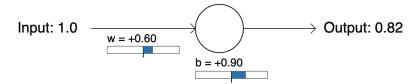
- (1.a) Use equidistant data points; (5 Points)
- (1.b) Read the note: "Runge's phenomenon" and use Chebyshev nodes. (10 Points)

In both cases, when you increase n, does your interpolant approximate the true function better? Show your results in plots, and explain why.

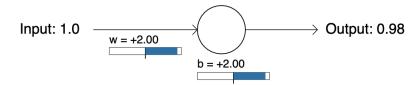
(2) (10 Points) Use linear piecewise interpolation. When you increase n, does your interpolant approximate the true function better? Show your results in plots, and explain why.

Problem 3: (20 Points) Use gradient descent to learn the weight and bias for a sigmoid neuron such that the neuron takes the input 1 to the output 0.

1) Initialize the weight and bias as the following:



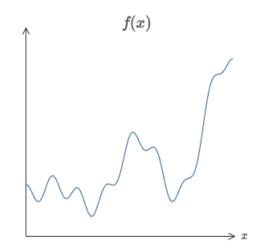
2) Initialize the weight and bias as the following:



For each case, learn up to 300 epochs. Plot the learning curve, i.e., the curve of cost vs. epoch. You need to determine an appropriate learning rate η .

Problem 4: (40 Points) Train a neural network so that for every possible input, $x \in [0,1]$, the output from the network g(x) is a close approximation for

$$f(x) = 0.2 + 0.4x^2 + 0.3x\sin(15x) + 0.05\cos(50x) .$$



You can use the python code (pytorch_regression.py) uploaded in Canvas as the starting point. And you need to install **PyTorch** from https://pytorch.org/.

Study the following aspects:

- 1. Determine an appropriate learning rate η ;
- 2. Determine appropriate mini-batch size and number of epoch;
- 3. Compare different activation functions (sigmoid, Tanh, and ReLU);
- 4. Compare different network architectures, e.g., with the same total number of neurons but different number of hidden layers, or with fixed number of layers but different numbers of neurons.