

Exercise Sheet 3

Due: 15.11.2023, 10:00

Download the file **poly.csv** from ISIS. The file consists of two columns. The first column contains input examples x_i and the second column contains the corresponding outputs y_i .

Exercise 3.1

Assume we are given a training set $(x_1, y_1), \dots, (x_m, y_m) \in \mathbb{R}^n \times \mathbb{R}$. Let $h_w(x) = w^\top \tilde{x}$, with $\tilde{x} = (1, x) \in \mathbb{R}^{n+1}$ and $w = (w_0, w_1, \dots, w_n) \in \mathbb{R}^{n+1}$. Consider the following empirical error function:

$$E_m[h_w, \lambda] = \frac{1}{2m} \sum_{i=1}^m (h_w(x_i) - y_i)^2 + \frac{\lambda}{2m} \sum_{j=1}^n w_j^2$$

Write the error function in matrix notation. Derive the update rule (in matrix notation) for the gradient descent method with respect to the parameter w .

Exercise 3.2

Implement the gradient descent method for polynomial regression in one variable with L_2 -regularization. Use matrix and vector operations instead of loops where possible.

Exercise 3.3

Apply the polynomial regression implemented in Exercise 3.2 to the poly-data. Dispense with the regularization here. Fit polynomials of order $k = 0, 1, \dots, 6$ and print the learned weights. Plot the MSE in dependence of the order k .

Exercise 3.4

Apply the polynomial regression with L_2 -regularization to the poly-data. For this, fit a polynomial of order $k=6$ and use the following regularization parameters:

$$\lambda \in \{0, 0.001, 0.003, 0.01, 0.03, 0.1, 0.3, 1, 3, 10\}.$$

Plot the MSE in dependence of the regularization parameter. For the sake of clarity, the x-axis should be scaled logarithmically. Create a plot showing the data points and for each $\lambda \in \{0, 0.01, 0.1, 1, 10\}$ the fitted polynomial. Moreover, print the learned weights for $\lambda \in \{0, 0.01, 0.1, 1, 10\}$. Discuss your results.

Exercise 3.5* (1 Bonus point)

Let $X \in \mathbb{R}^{m \times n+1}$ be an augmented data matrix and $y \in \mathbb{R}^m$ the output vector.

- Show that Multiple Linear Regression has a unique solution if $\text{rank}(X) = n + 1$.
- State a necessary condition on m for having a unique solution.
- What if $\text{rank}(X) < n + 1$? Does a solution exist? Is the solution unique? Does it depend on y ? Prove your claims.

To solve this task, you can use the following basic results from linear algebra:

- 1) $\text{rank}(X) = \text{rank}(X^\top X) = \text{rank}(X^\top)$
- 2) $\text{im}(X^\top) = \text{im}(X^\top X)$
- 3) A system of linear equations $Ax = b$ with $A \in \mathbb{R}^{n \times n}$, $b \in \mathbb{R}^n$ has
 - a) a unique solution if $\text{rank}(A) = \text{rank}([A|b]) = n$
 - b) infinitely many solutions if $\text{rank}(A) = \text{rank}([A|b]) < n$
 - c) no solution if $\text{rank}(A) < \text{rank}([A|b])$,
 where $[A|b] \in \mathbb{R}^{n \times n+1}$ is the concatenated matrix composed of A and b .