



Exercise Sheet 4

Machine Learning Basics

Deadline: 26.11.2018, 23:59

Exercise 4.1 - Gradient Descent and Newton's Method (1.5 + 1.5 + 0.5 + 0.5 = 4 points)

In the optimization setting, Gradient Descent and Newton's Method are two commonly used iterative methods for finding a solution. Let's say we want to minimize a function $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ which is defined as $f(\mathbf{x}) = x_1^2 - 3x_1 + x_2^2 - x_1x_2$. The starting point is $\mathbf{x}^{(0)} = [1, 1]^T$.

- Use Gradient Descent with step size (also known as learning rate) $\epsilon = 0.5$ to minimize the function f . The iteration should stop if the L2-norm of the gradient at the current point is less than 0.2. Show your intermediate steps. [Hint: The iteration should finish within 4 steps].
- Starting from the same point \mathbf{x} , use Newton's Method to find a solution. Is the solution you get a global minimum? Argue why or why not.
- We learned from the previous Exercise 3.3 that a bad step size can lead to non-local minimum for Gradient Descent. But are we guaranteed to converge to a local minimum when using Newton's Method since it does not have an explicit step size? What is the implicit step size used in Newton's Method?
- Is Newton's Method always applicable if the function f is twice continuously differentiable? Argue with the help of the function $f(x) = 2x^3 - 5x$ at $x = 0$.

Exercise 4.2 - Overfitting

(0.5 + 1.5 + 1 = 3 points)

Overfitting happens when the model capacity is too high and there is no proper regularization applied. Figure 1 and Figure 2 show two different classification boundaries for a binary classification problem, where the blue points and the red points represent the training data consisting out of two classes. Please answer the following questions.

- Which classification boundary correspond to the overfitting and the underfitting, respectively?
- Explain the terms overfitting, underfitting, and model capacity with the help of Figure 1 and Figure 2.
- What happens to the training error and validation error when a model overfits? Explain.

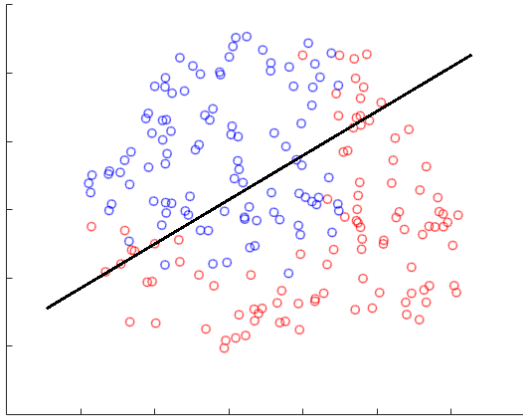


Figure 1

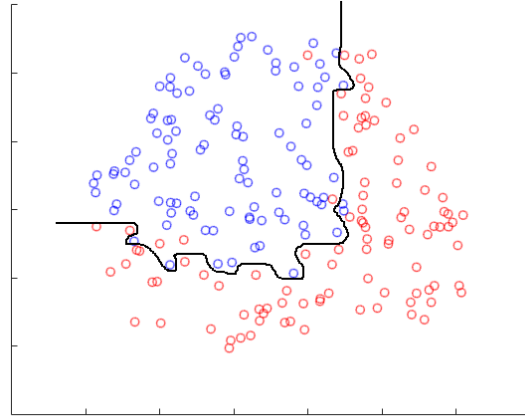


Figure 2

Exercise 4.3 - Regularization

(1 + 1 + 1 = 3 points)

The solution of the linear regression problem given on slide 14 of chap5 is also known as *normal equation*. However, the inverse of $\mathbf{x}^T \mathbf{x}$, where \mathbf{x} is the design matrix, may not exist. Therefore you may have to solve a linear system rather than applying the closed form solution directly. We saw in the lecture a modified linear regression problem, where we minimize a loss function $J(\mathbf{w}) = MSE_{train} + \lambda \mathbf{w}^T \mathbf{w}$ for \mathbf{w} , $\lambda > 0$. This modified version is known as *ridge regression*. One of the purposes of ridge regression is to obtain a unique solution for the normal equation when $\mathbf{x}^T \mathbf{x}$ is not invertible.

- Given the same setting of linear regression from the lecture (see slide 14, chap5), derive the closed form solution for the ridge regression. [Hint: you may want to use some formulas from Exercise 3.1.]
- Argue (no need for rigorous proof) for the uniqueness of the closed form solution for the ridge regression, i.e. argue why the linear system has a unique solution.
- Argue (no need for rigorous proof) that the solution you get from a) is indeed a global minimizer.

Submission instructions

The following instructions are mandatory. If you are not following them, tutors can decide to not correct your exercise.

- You **have to** submit the solutions of this assignment sheet as a team of 2-3 students.
- Hand in a **single** PDF file with your solutions.
- Therefore make sure to write the student ID and the name of each member of your team on your submission.
- Your assignment solution must be uploaded by only **one** of your team members to the course website.
- If you have any trouble with the submission, contact your tutor **before** the deadline.
- Don't forget to attach a solution of exercise 3.2 (if not already submitted).**