# Exercise 1 - Math Refresher

Winter term 2019/2020 name1, name2, name3



### **Important**

Please solve the assignments in groups of 3 to 4 students. The solutions are going to be presented and discussed after the submission deadline. Sample solutions will not be uploaded. However, you are free to share correct solutions with your colleagues **after they have been graded**. Please submit your solutions via Moodle **and** in printed form. Only one member of the group has to submit the solutions. Therefore, make sure to specify the names of all group members. Please do not submit hand-written solutions, rather use proper type-setting software like LATEX or other comparable programs.

Your homework will be corrected and given back to you. Correct solutions are rewarded with a bonus for the exam (max. 10 percent, if all solutions submitted are correct). Please note: You have to pass the exam without the bonus points! (i.e. it is not possible to turn 5.0 into 4.0) The solutions have to be your own work. If you plagiarize, you will lose all bonus points!

#### Further remarks:

- Code assignments have to be done in Python
- The following packages are allowed: numpy, pandas (please ask, if you want to use a specific package not mentioned here)
- Do not use already implemented models (e.g. from scikit-learn)



# 1 Linear Algebra Refresher

a) Matrix Operations (1 point)

A fellow student suggests that matrix addition and multiplication are very similar to scalar addition and multiplication, i. e. commutative, associative and distributive. Is this a correct statement? Prove it mathematically or disprove it by providing at least one counter example per property (commutativity, associativity, distributivity).

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## b) Matrix Inverse (1 point)

What is a matrix inverse? How can you build the inverse of a non-square matrix? You would like to invert a matrix  $M \in \mathbb{R}^{2 \times 3}$ , write down the equation for computing it and specify the dimensionality of the matrices after each single operation (e. g. multiplication, inverse).



c) Eigenvectors and Eigenvalues (1 point)

Explain what eigenvectors and eigenvalues of a matrix  $\boldsymbol{A}$  are. Why are they relevant in machine learning?



## 2 Statistics Refresher

a) Terminology (1 point)

What is a random variable? What is a probability density function (PDF)? What is a probability mass function (PMF)? What do a PDF and a PMF tell us about a random variable?

Solution:

b) Expectation and Variance (1 point)

State the general definition of expectation and variance for the probability density  $f:\Omega\to\mathbb{R}$  of a continuous random variable. What do expectation and variance express?



# 3 Optimization

a) Numerical Optimization - Gradient Descent (5 points)

Implement a simple gradient descent algorithm for finding a minimum of the Rosenbrock function with n=2 using Python and NumPy:

$$f(\mathbf{x}) = \sum_{i=1}^{n-1} \left[ 100(x_{i+1} - x_i^2)^2 + (x_i - 1)^2 \right]$$

Submit your code and a plot of the learning curve for the best run of your gradient descent implementation. Which learning rate worked best? (Hint: You need to find the first derivative(s) of f(x) for n=2 and iteratively evaluate them during gradient descent. Automatic differentiation tools are not allowed for this exercise. Choose a random starting point for the parameters, for example  $x \in [-2, 2]$ .)