

# Calculus and Linear Algebra II

## Homework 4

Due on April 20, 2020

### Problem 1 [8 points]

Find the solution to the logistic growth model

$$\frac{dy}{dx} = \lambda y \left(1 - \frac{y}{k}\right)$$

by separation of variables, where  $\lambda, k > 0$  are some parameters.

### Problem 2 [4 points]

The point of this exercise is to give an example of an ODE where a solution exists but it is not unique. Show that this is the case for the ODE

$$\frac{dy}{dx} = y^{1/3},$$

for initial condition  $y(0) = 0$ . Specifically, show that there exist at least two solutions that satisfy the initial condition. Can you even give a whole family of solutions  $y_\lambda(x)$ , with each solution different for different  $\lambda \in \mathbb{R}$ ?

### Problem 3 [6 points]

We consider the ODE

$$\frac{dy}{dx} = y^2 + \lambda y + 1,$$

for some parameter  $\lambda \in \mathbb{R}$ . Without solving the ODE explicitly discuss the qualitative behavior of the solutions for different values of the parameter  $\lambda$  and for different initial conditions. (I.e., what are the equilibrium positions, are they stable or not, how do the solutions behave for large  $x$ ?)

### Problem 4 [4 points]: The Curse of Dimension

Let us consider the standard basis  $\vec{e}_1, \dots, \vec{e}_n$  of  $\mathbb{R}^n$ .

- Compute the unit vector (= vector with length one) in the direction of the diagonal  $\{\lambda(1, \dots, 1) : \lambda \in \mathbb{R}\}$ .
- Show that this unit vector is, in the limit  $n \rightarrow \infty$ , orthogonal to any basis vector  $\vec{e}_j$ ,  $j = 1, \dots, n$ .

*Note: So with finite precision, they are actually orthogonal to each other for very large  $n$ , which contradicts our intuition from  $\mathbb{R}^2$  or  $\mathbb{R}^3$ . This creates lots of problems for data mining, e.g., when performing a so-called Principal Component Analysis.*

**Problem 5 [8 points]: A least norm problem: Least-squares solutions of linear equations**

- (a) Generalize the method of Lagrange multipliers to the case where  $m$  constraints  $G_j(\vec{x}) = 0$ ,  $j = 1, \dots, m$  are present.
- (b) Let us consider the following example of a so-called convex optimization problem (which are very relevant for machine learning). We aim at minimizing the function  $f(\vec{x}) = |\vec{x}|^2 = \sum_{i=1}^n x_i^2$ , subject to the constraint  $\vec{G}(\vec{x}) = A\vec{x} - \vec{b} = \vec{0}$ , where  $\vec{x} \in \mathbb{R}^n$ ,  $\vec{b} \in \mathbb{R}^m$ ,  $\vec{G}(\vec{x}) \in \mathbb{R}^m$ , and  $A$  a real  $m \times n$  matrix with  $\text{rank} A = m < n$ . For example,  $A$  could come from some data set, and we want to find the solution to the system of linear equations with the smallest norm  $|\vec{x}|^2$ . Note that the matrix  $A$  has rank  $m < n$ , so it is not invertible. Hence, we cannot simply insert the constraint into  $f$  and compute the minimum. But luckily we know about Lagrange multipliers. So find the minimum of  $f$  under the  $m$  constraints  $\vec{G}(\vec{x}) = \vec{0}$ . *Hint: The matrix  $AA^T$  is invertible (why?), where  $A^T$  is the transpose matrix with entries  $(A^T)_{ij} = A_{ji}$ , i.e., where rows and columns are interchanged.*

**Bonus Problem [8 points]**

The equation of motion of a particle with charge  $q > 0$  and mass  $m > 0$  in an electric field  $\vec{E}$  and a magnetic field  $\vec{B}$  is

$$m \frac{d^2 \vec{x}(t)}{dt^2} = q \left( \left( \vec{x}'(t) \times \vec{B} \right) + \vec{E} \right),$$

where  $\vec{x}(t) \in \mathbb{R}^3$  is the trajectory, and  $\vec{x}'(t)$  the velocity of the particle. (This is the Lorentz force law.) Let us now consider the particular electric field

$$\vec{E} = (0, 0, E_3), \quad \text{with } E_3 > 0,$$

and magnetic field

$$\vec{B} = (B_1, 0, 0), \quad \text{with } B_1 > 0.$$

Find the solution to the equation of motion for initial data

$$\vec{x}(0) = \vec{0}, \quad \vec{x}'(0) = (0, v_0, 0), \quad \text{with } v_0 \geq 0.$$

Then, discuss the qualitative behavior of the solution depending on the parameter

$$\lambda = \frac{v_0}{E_3/B_1}.$$