

# Optimization and Computational Linear Algebra for Data Science

## Homework 10: Regression

Due on November 26, 2019

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- Unless otherwise stated, all answers must be mathematically justified.
  - Partial answers will be graded.
  - You can work in groups but each student must write his/her own solution based on his/her own understanding of the problem. Please list on your submission the students you work with for the homework (this will not affect your grade).
  - Problems with a  $(\star)$  are extra credit, they will not (directly) contribute to your score of this homework. However, for every 4 extra credit questions successfully answered your lowest homework score get replaced by a perfect score.
  - If you have any questions, feel free to contact me ([1m4271@nyu.edu](mailto:1m4271@nyu.edu)) or to stop at the office hours.
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DRAFT

**Problem 10.1** (2 points). *Let  $A \in \mathbb{R}^{n \times m}$ . Show that if  $A$  has linearly independent columns, then  $A^\dagger = (A^\top A)^{-1} A^\top$ .*

**Problem 10.2** (2 points). *Let  $A \in \mathbb{R}^{n \times d}$  and  $y \in \mathbb{R}^n$ . The Ridge regression adds a  $\ell_2$  penalty to the least square problem in order to “select” a solution of smaller norm. The Ridge regression problem is then:*

$$\text{minimize} \quad \|Ax - y\|^2 + \lambda \|x\|^2 \quad \text{with respect to } x \in \mathbb{R}^d, \quad (1)$$

*for some penalization parameter  $\lambda > 0$ . Show that (1) admits a unique solution given by*

$$x^{\text{Ridge}} = (A^\top A + \lambda \text{Id})^{-1} A^\top y.$$

**Problem 10.3** (3 points).

**Problem 10.4** (3 points).

**Problem 10.5** (★).

