

# Comparison of gradient schemes

## 1 Gradient descent

Consider the following least square problem

$$\min_{x \in \mathbb{R}^n} \{F(x) \stackrel{\text{def}}{=} \frac{1}{2} \|Ax - b\|^2\}, \quad (1.1)$$

where  $b \in \mathbb{R}^n$  and  $A \in \mathbb{R}^{n \times n}$  is of the form

$$A = \begin{bmatrix} 2 & -1 & & & \\ -1 & 2 & -1 & & \\ & & \ddots & & \\ & & -1 & 2 & -1 \\ & & & -1 & 2 \end{bmatrix}_n.$$

**Numerical setting** Size of the problem  $n = 201$ , under this setting the condition number of the problem is roughly  $\text{cnd} = 2.7 \times 10^8$ . Vector  $b$  is a Gaussian random vector.

**Algorithms** Solve (1.1) with gradient descent, heavy-ball method and Nesterov's optimal scheme, compare the performance of these schemes. In terms of objective function value and  $\|x_k - x_{k-1}\|$ .

## 2 Proximal gradient descent

Consider the  $\ell_1$ -regularised least square problem, which is also known as LASSO problem.

$$\min_{x \in \mathbb{R}^n} \mu \|x\|_1 + \frac{1}{2} \|Ax - b\|^2. \quad (2.1)$$

This is a typical problem that can be handled by proximal gradient descent method.

**Data** The **australian** data set from LIBSVM<sup>1</sup> is considered. See also the download link on course website.

**Algorithms** Solve (2.1) with proximal gradient descent, proximal version of heavy-ball method and FISTA scheme, compare the performance of these schemes. In terms of objective function value and  $\|x_k - x_{k-1}\|$ .

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<sup>1</sup><https://www.csie.ntu.edu.tw/~cjlin/libsvmtools/datasets/>