

# IE510 Applied Nonlinear Programming Homework 3

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Date assigned: Week 10 Monday 03/19/2018

Date due: Week 12 Tuesday 04/03/2018, at 3pm

Remark: The work you are submitting for this homework assignment must be your own, and suspiciously similar homework submitted by multiple individuals may be reported to the University for investigation.

## 1 Reading

- Reading: Textbook Section 1.5.2 “Incremental Gradient Method”
- Reading: Textbook Section 2.7 “Block Coordinate Descent Methods” (note that the theorem in 1999 version needs to be modified; see the correction in the Box file)
- If interested in more details about incremental gradient methods and SGD, you can read references listed in lecture slides

## 2 Problems

1. (60 points + 5 bonus points) Consider a non-singular matrix  $A = (A_1, A_2, \dots, A_n)$ , where  $A_i \in \mathbb{R}^{d \times 1}$  and  $\|A_i\| = 1$ , for  $\forall i$ . We want to solve  $\min_x \|Ax - b\|^2$ , s.t.  $x \in \mathbb{R}^n$ . Assume  $b = 0$  for simplicity.

(a) (25 points) Let  $n = 100$ . Consider the setting  $A_{ij} \sim \text{Unif}[0, 1]$ , where  $\text{Unif}[0, 1]$  represents uniform distribution on the interval  $[0, 1]$ . After generating  $A$ , normalize  $A_i$ , s.t.  $\|A_i\| = 1$ ,  $\forall i$  (i.e. define  $\tilde{A}_{ij} = \frac{A_{ij}}{\|A_i\|}$ ,  $\forall i, j$ , and replace  $A$  by  $\tilde{A}$ ).

Use random initial point  $x^0 = (x_1^0, \dots, x_n^0)$ , where  $x_i^0 \sim \mathcal{N}(0, 1)$ ,  $i = 1, \dots, n$ . Here  $\mathcal{N}(0, 1)$  represents standard Gaussian distribution. Compare two algorithms randomized coordinate descent (R-CD) and gradient descent (GD) with stepsize  $\frac{1}{L}$  by plotting function values v.s. epochs.

Remark: One epoch of R-CD consists of  $n$  iterations.

(b) (5 points) I claimed in class that R-CD is roughly  $\frac{\lambda_{\max}(A^T A)}{\lambda_{\text{avg}}(A^T A)}$  times faster than GD, where  $\lambda_{\text{avg}}(A^T A)$  is the average eigenvalue of  $A^T A$ . Do your figures in (a) support this claim?

(c) (30 points) Under the same setting as (a), compare SGD with various constant stepsizes, SGD with diminishing stepsizes and GD with stepsize  $\frac{1}{L}$ . Report and discuss your findings.

(d) (5 bonus points) Generate  $Q$  as  $Q_{ii} = 1, \forall i; Q_{ij} = c, \forall i \neq j$ . Here  $0 \leq c \leq 1$  is a constant. Consider problem

$$\min_x x^T Q x, \quad s.t. \quad x \in \mathbb{R}^n.$$

Compare 4 algorithms: cyclic CD(C-CD), randomized CD(R-CD), randomly permuted CD(RP-CD), GD with stepsize  $\frac{1}{L}$ . Run experiments for different  $n$  and  $c$ . What do you find? Can you make some general conjectures that explain your findings?

2. (40 points) Consider the following problem:

$$\min_{x \in \mathbb{R}^n} f(x) = \frac{1}{2} \|M - xx^T\|_F^2. \quad (*)$$

(a) (5 points) Describe how to use gradient descent to solve (\*).

(b) (10 points) Describe how to use coordinate gradient descent to solve (\*).

(c) (15 points) Describe how to use coordinate descent to solve (\*).

Note: You can assume there exists a 3-rd order equation solver that returns all three complex roots of equation  $az^3 + bz^2 + cz + d = 0$ .

(d) (10 points) Describe how to use stochastic gradient method to solve (\*). Please provide at least two versions of SGD.

Remark: You can use any update order in all algorithms.