## Question 1: ADMM (30 points)

Solve the following optimization problem using the scaled form of alternating direction method of multipliers (ADMM).

$$\min_{x} \frac{1}{2} x^T P x + q^T x + \frac{\lambda}{2} ||z||_2^2 \qquad s.t. \qquad \begin{cases} x = z \\ a \le z \le b \end{cases}$$

Where  $P \in \mathbb{R}^{n \times n}$  and,  $a, b, x, q \in \mathbb{R}^n$ .

**Part1.** Write the augmented Lagrangian function (the scaled form) and drive the ADMM updates (Show your work).

**Part 2.** "Question1.mat" dataset contains variables P, q, a and b. Set  $\lambda = 0.5$  and  $\rho = 1.1$ . Plot the value of  $\frac{1}{2}x^TPx + q^Tx$  per iteration. Plot  $||x - z||_2$  per iteration. Plot your final value of x.

## **Question 2: Coordinate Descent (35 points)**

Part 1: Solve the following optimization problem using coordinate descent algorithm.

$$\min_{w} \frac{1}{2} \|y - Xw\|_{2}^{2} + \frac{\lambda}{2} \|w\|_{2}^{2}$$

Where  $y, w \in \mathbb{R}^n$  and  $X \in \mathbb{R}^{n \times n}$ . Consider  $\lambda = 0.1$ . Drive a closed form solution for w.

Part 2: Solve the following optimization problem using coordinate descent algorithm.

$$\min_{w} \frac{1}{2} ||y - Xw||_{2}^{2} + \lambda_{1} |w|_{1} + \frac{\lambda_{2}}{2} ||w||_{2}^{2}$$

Drive a closed form solution for w. Consider  $\lambda_1 = 0.05$  and  $\lambda_2 = 0.01$ .

**Part 3.** In this part, you implement your own regression algorithm using your solution in part 1 and 2 to predict the performance decay over time of the Gas Turbine (GT) compressor. The range of decay of compressor has been sampled with a uniform grid of precision 0.001. The compressor decay coefficient is in the range of [0.95,1]. The dataset is provided as "Question2.csv". The last column of the datasets corresponds to the output we want to predict.

## - The 13 features are:

Lever position (lp)

Ship speed (v) [knots]

Gas Turbine (GT) shaft torque (GTT) [kN m]

GT rate of revolutions (GTn) [rpm]

Gas Generator rate of revolutions (GGn) [rpm]

Port Propeller Torque (Tp) [kN]

Hight Pressure (HP) Turbine exit temperature (T48) [C]

GT Compressor outlet air temperature (T2) [C]

HP Turbine exit pressure (P48) [bar]

GT Compressor outlet air pressure (P2) [bar]

GT exhaust gas pressure (Pexh) [bar]

Turbine Injection Control (TIC) [%] Fuel flow (mf) [kg/s] GT Compressor decay state coefficient

- Use the first 2000 samples to learn  $w_{opt}$ . Report your coefficients  $(w_{opt})$  for each method. Plot the function  $||y Xw||_2^2$  versus iterations.
- Report your Sum of Absolute Errors on the test set (data samples 2001-2387) for each method.

## **Question 3: Proximal Gradient Descent (35 points)**

Solve the following optimization problem using proximal gradient method.

$$f(\boldsymbol{\theta}) = \min_{\boldsymbol{\theta} \in \mathbb{R}^d} \frac{1}{m} \sum_{i=1}^m \left[ \log \left( 1 + \exp(\boldsymbol{x}_i \boldsymbol{\theta}) \right) - y_i \boldsymbol{x}_i \boldsymbol{\theta} \right] + \frac{\lambda_2}{2} \|\boldsymbol{\theta}\|_2^2 + \lambda_1 \|\boldsymbol{\theta}\|_1$$

Where  $\log(\cdot)$  is the natural logarithm;  $\mathbf{x}_i \in \mathbb{R}^{1 \times d}$  is sample i;  $\mathbf{\theta} \in \mathbb{R}^d$ ,  $y_i \in \{0,1\}$  is label for sample i and  $\lambda > 0$ .

Decomposed  $f(\theta)$  into a convex and differentiable function g and a convex but not differentiable function h.

Part 1: Derive the proximal gradient method updates for the objective function. (Show your work)

**Part 2:** "Question3.csv" contains 2000 data. There are two classes "0" and "1". The last column represents the labels. Use the first 1000 data samples to learn  $\theta$ , and the rest to evaluate the accuracy of your classifier. Use your learned  $\theta$  to predict the labels on the test set (Predict labels =  $\frac{1}{1 + \exp(-x\theta)}$ ). Set  $\lambda_1 = 10$  and  $\lambda_2 = 5$ . Build your own classifier using your equations (part 1).

- Plot accuracy per iteration on the training set.
- Report  $\|\boldsymbol{\theta}\|_0$ .
- Report your classification accuracy on the test set.