Midterm written test (20 points)

This assignment is worth 20 points. Students do not need to complete all the tasks.

- 1. (3 points) Show that the following set is convex $D = \{x : (\mathbf{x}^T \mathbf{Q} \mathbf{x} + 1)^2 + \ln \sum_i \exp(x_i) \}$, where $\mathbf{Q} > 0$.
- 2. (3 points) Find the optimal solution of the problem

$$\max_{x=[x_1,x_2]} (x_1-2)^2 + 2(x_2-3)^2$$
s.t. $-2x_1 + x_2 \le 0$ (2)

$$s.t. -2x_1 + x_2 \le 0 (2)$$

$$x_1 + x_2 \le 4 \tag{3}$$

$$x_2 \ge 0 \tag{4}$$

3. (3 points) Derive the solution for the following optimization problem

$$\min_{\mathbf{x} \in \mathbb{R}^n} \quad \|\mathbf{b} - \mathbf{x}\|_2^2 \tag{5}$$

s.t
$$x \ge 0$$
 (6)

4. (3 points) Derive the optimality conditions and solution for the following optimization problem

$$\min_{\boldsymbol{x} \in \mathbb{R}^n} \|\boldsymbol{b} - \boldsymbol{x}\|_2^2 \tag{7}$$

s.t
$$x \ge 0$$
 (8)

$$\mathbf{1}^T \mathbf{x} \le 1 \tag{9}$$

where $\boldsymbol{b} \in \mathbb{R}^n_+$.

5. (3 points) Derive the optimal solution for the following optimization problem

$$\min_{\mathbf{x} \in \mathbb{R}^n} \|\mathbf{b} - \mathbf{x}\|_2^2 + \gamma \|\mathbf{x}\|_1$$
 (10)

6. (10 points)

• (7 points)

Derive optimal solution and implement algorithm for the following optimization problem

$$\min_{\mathbf{X} \in \mathbb{R}^{n \times m}} \|\mathbf{Y} - \mathbf{X}\|_F^2 \tag{11}$$

s.t
$$\|\mathbf{X}\|_* \le \delta$$
 (12)

where **Y** is a matrix of size $n \times m$ and $\delta > 0$.

• (3 points) Derive optimal solution and implement algorithm for the following optimization problem

$$\min_{\mathbf{X}_{1}, \mathbf{X}_{2}} \|\mathbf{Y}_{1} - \mathbf{X}_{1}\|_{F}^{2} + \|\mathbf{Y}_{2} - \mathbf{X}_{2}\|_{F}^{2}
\text{s.t.} \quad \gamma_{1} \|\mathbf{X}_{1}\|_{*} + \gamma_{2} \|\mathbf{X}_{2}\|_{*} \leq \delta$$
(13)

s.t
$$\gamma_1 ||\mathbf{X}_1||_* + \gamma_2 ||\mathbf{X}_2||_* \le \delta$$
 (14)

where \mathbf{Y}_1 and \mathbf{Y}_2 are matrices of size $n \times m$ and $\delta > 0$.

7. (10 points) Consider Support Matrix Machine problem which finds a hyperplane {X | $tr(\mathbf{W}^T\mathbf{X}) + b = 0$ } such that **W** has minimal rank

$$\begin{aligned} & \min_{\mathbf{W},b} & \frac{1}{2} ||\mathbf{W}||_F^2 + \gamma ||\mathbf{W}||_* \\ & \text{s.t.} & y_i(\text{tr}(\mathbf{W}^T \mathbf{X}_i) + b) \geq 1, \quad i = 1, 2, \dots, L. \end{aligned}$$

where X_i are feature matrices of size $n \times m$ and the labels $y_i \in \{-1, 1\}$ are for the samples $i = 1, \dots, L, \gamma > 0$. $\|\mathbf{W}\|_*$ represents the nuclear norm of \mathbf{W} .

- Develop an algorithm based on the ADMM method to find **W** and *b*,
- Show examples for training images, e.g., for the two digits 0 and 1 in the MNIST dataset
- Report accuracy of the classifier and ranks of W for different values of the regularizer parameter γ