

# Homework requirements

- Digital format (can be typeset or photos) is preferred
- Submit this homework and ALL past homeworks by 23:59 of next Wednesday (July 24<sup>th</sup>).
- Each homework 10 points; 1 point deducted for each day of delay

# Contact information

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# Homework (6)

1. [Equality constrained entropy maximization.] We consider the equality constrained entropy maximization problem

$$\min_{\mathbf{x}} f(\mathbf{x}) = \sum_{i=1}^n x_i \log x_i, \quad s.t. \quad \mathbf{A}\mathbf{x} = \mathbf{b}.$$

Use the following methods to solve it at  $p = 100$  and  $n = 500$  (you may generate  $\mathbf{A}$  and  $\mathbf{b}$  randomly):

- Direct projected gradient, with inexact line search.
- Dual approach.

Write a report to describe your settings and compare their performance (numerical accuracy vs. iteration number). Codes should also be handed in.

# Homework (6)

2. Consider the problem

$$\min_{\mathbf{x}} \frac{1}{2} \|\mathbf{x}\|_2, \quad s.t. \quad \mathbf{A}\mathbf{x} = \mathbf{b}, \quad (1)$$

where  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\mathbf{b} \in \mathbb{R}^m$ ,  $m \leq n$ , and  $\text{rank } \mathbf{A} = m$ .

Randomly generate  $\mathbf{A} \in \mathbb{R}^{200 \times 300}$  and  $\mathbf{b} \in \mathbb{R}^{200}$  and solve problem (1) numerically by the penalty method. Hand in your code and report.

3. Consider:

$$\min_{\mathbf{x} \in \mathbb{R}^N} \|\mathbf{y} - \mathbf{D}\mathbf{x}\|_2^2, \quad s.t. \quad \|\mathbf{x}\|_? \leq 1,$$

where  $\|\mathbf{x}\|_?$  can be either the  $\ell_1$  norm or the  $\ell_\infty$  norm. Randomly generate  $\mathbf{D} \in \mathbb{R}^{200 \times 300}$  and  $\mathbf{y} \in \mathbb{R}^{200}$  and use Frank-Wolfe algorithm to solve it, for both  $\ell_1$  norm and  $\ell_\infty$  norm. Further compare F-W algorithm with the projected gradient descent in convergence speed (objective function value vs. iteration number). Hand in your code and report.

# Homework (6)

4. Use LADMPSAP to solve a graph construction problem:

$$\min_{\mathbf{Z}, \mathbf{E}} \|\mathbf{Z}\|_* + \lambda \|\mathbf{E}\|_{2,1}, \quad \text{s.t.} \quad \mathbf{D} = \mathbf{DZ} + \mathbf{E}, \mathbf{Z}^T \mathbf{1} = \mathbf{1}, \mathbf{Z} \geq \mathbf{0}, \quad (1)$$

where  $\mathbf{1}$  is an all-one vector. Randomly generate  $\mathbf{D} \in \mathbb{R}^{200 \times 300}$ . Hand in your code and report.

# Homework (6)

5. Use block coordinate descent to solve the low-rank matrix completion problem:

$$\min_{\mathbf{U}, \mathbf{V}, \mathbf{A}} \frac{1}{2} \|\mathbf{UV}^T - \mathbf{A}\|_F^2, \quad s.t. \quad \mathcal{P}_\Omega(\mathbf{A}) = \mathcal{P}_\Omega(\mathbf{D}),$$

where  $\mathcal{P}_\Omega(\cdot)$  is an operator that extracts entries of a matrix whose indices are in  $\Omega$  and sets the remaining entries zeros.

Randomly generate  $\mathbf{D} = \mathbf{U}_0 \mathbf{V}_0^T$  and  $\Omega$ , where  $\mathbf{U}_0 \in \mathbb{R}^{200 \times 5}$ ,  $\mathbf{V}_0 \in \mathbb{R}^{300 \times 5}$  and  $|\Omega| = 0.1 \times 200 \times 300$ . Hand in your code and report showing your settings and the difference  $\|\mathbf{A}^* - \mathbf{D}\|_2$  where  $\mathbf{A}^*$  is the optimal solution.