

ENMG 616 Advanced Optimization & Techniques

Homework Assignment





Due Date: November 15, 2023


*You need to show the steps followed to get the final answer (Do not just give the final result). The homework should be submitted to Moodle as **one** pdf file. Please insert a copy of your Matlab code in the submitted file.*

1 Solving Quadratic Constrained Optimization


Consider the following quadratic optimization problem

$$\min_{\mathbf{x} \in \mathbb{R}^n} \frac{1}{2} \mathbf{x}^T \mathbf{Q} \mathbf{x} + \mathbf{q}^T \mathbf{x} \quad \text{s.t.} \quad 1 \leq \mathbf{x} \leq 3$$

1. Randomly generate \mathbf{Q} and \mathbf{q} according to the following procedure:
 - (a) Set $n = 100$;
 - (b) Generate a random matrix $\mathbf{B} \in \mathbb{R}^{n \times n}$ where each element is drawn from standard normal distribution. Matlab Function: $\mathbf{B} = \text{randn}(n, n)$.
 - (c) Let $\mathbf{Q} = \mathbf{B} * \mathbf{B}'$.
 - (d) For \mathbf{Q} to be a PSD matrix, add its smallest eigenvalue plus 10. 
Matlab Function: $\mathbf{Q} = \mathbf{Q} + (\text{MinEig} + 10) * \text{eye}(n)$.
 - (e) Generate a random vector \mathbf{q} where each element is drawn from normal distribution $\mathcal{N}(0, 100)$. Matlab Function: $\mathbf{q} = 10 * \text{randn}(n, 1)$.
2. Compute the Lipschitz constant L of the objective function. 
3. Implement the gradient projection method with constant step-size $\alpha_r = \frac{1}{L}$. 
4. Implement the gradient projection method with diminishing step-size $\alpha_r = \frac{5 \ln(r)}{rL}$. 

!! 5. Implement Frank-Wolfe method. 

6. Run the above 3 algorithms from the point $\mathbf{x}^0 = 1$ for 2000 iterations, and plot the following error

$$\mathcal{E}(\mathbf{x}^r) = \|\mathbf{x}^r - [\mathbf{x}^r - \nabla f(\mathbf{x}^r)]_+\|_2^2$$


versus iteration number r .

1.1 Dual Problem

1. Write down the Lagrangian function (Hint you can split the constraints into $\mathbf{x} \leq 3$ and $\mathbf{x} \geq 1$). ✓
2. Find the dual problem. ✓
3. Write the K.K.T conditions. ✓
4. Implement and run the gradient projection method with constant step-size for the primal and dual problem to reach convergence. ✓
5. Conclude that strong duality holds? ✓

1.2 Duality

Repeat the above experiment (only running the codes) when Q is generated as follows: $\mathbf{B} = \text{randn}(n, n)$; $\mathbf{Q} = \frac{1}{2}(\mathbf{B} + \mathbf{B}^T)$. Comment on the results.

2 Projection to a set of linear equality constraints

Consider the following optimization problem

$$\min_{\mathbf{x} \in \mathbb{R}^n} \frac{1}{2} \|\mathbf{z} - \mathbf{x}\|_2^2 \quad \text{s.t. } \mathbf{Ax} = \mathbf{b}.$$

This problem aims at finding the projection of \mathbf{z} to the set $\{\mathbf{x} \mid \mathbf{Ax} = \mathbf{b}\}$. Assuming that \mathbf{AA}^T is invertible.

1. Write down the Lagrangian function. ✓
2. Find the dual problem. ✓
3. Write down the KKT conditions. ✓
4. Using KKT conditions express the optimal solution \mathbf{x}^* as a function of \mathbf{z} , \mathbf{A} , and \mathbf{b} . ✓