

Homework Assignment 8: Due Wednesday, May 29

Problem 1. Consider the following convex optimization

$$\begin{array}{ll}\min & (x_1 - 1)^2 + (x_2 + 1)^2 \\ \text{s.t.} & -x_1 + x_2 - 1 \geq 0.\end{array}$$

Show the following statements:

- (a) Write out the necessary and sufficient optimality condition and find an optimal solution for this problem.
- (b) Write out its Wolfe dual and find an optimal solution for the dual problem.
- (c) Analyze the relationship between the primal and dual problems.

Problem 2. Consider the problem

$$\text{minimize } 5x^2 + 5y^2 - xy - 11x + 11y + 11.$$

- (a) Find a point satisfying the first-order necessary conditions for a solution.
- (b) Show that this point is a global minimum.
- (c) What would be the rate of convergence of steepest descent for this problem?
- (d) Starting at $x = y = 0$, how many steepest descent iterations would it take (at most) to reduce the function value to 10^{-11} ?

Problem 3. Suppose we use the method of steepest descent to minimize the quadratic function $f(x) = \frac{1}{2}(x - x^*)^T Q(x - x^*)$ where Q is a symmetric and positive definite matrix

and x^* is the optimal solution, but we allow a tolerance $\pm\delta\alpha_k$ ($\delta \geq 0$) in the line search, that is

$$x_{k+1} = x_k - \alpha_k g_k \quad \text{with} \quad g_k = \nabla f(x_k),$$

where

$$(1 - \delta)\bar{\alpha}_k \leq \alpha_k \leq (1 + \delta)\bar{\alpha}_k$$

and $\bar{\alpha}_k$ minimizes $f(x_k - \alpha g_k)$ over α .

(a) Find the convergence rate of the algorithm in terms of the smallest and largest eigenvalues of Q and the tolerance δ . (*Hint: Assume the extreme case $\alpha_k = (1 + \delta)\bar{\alpha}_k$.*)

(b) What is the large δ that guarantees convergence of the algorithm? Explain this result geometrically.

(c) Does the sign of δ make any difference?

Problem 4. Using Newton's method find a zero and then a minimum or maximum for $f(x) = \sin(x) - (x/2)^2$ with an initial point $x_0 = 1.5$. Write out the process of iteration.

Problem 5. Write a program (MatLab or C) of the method of steepest descent to solve the problem

$$\text{minimize} \quad f(x) = \frac{1}{2}x^T Qx - b^T x,$$

where

$$Q = \begin{pmatrix} 0.78 & -0.02 & -0.12 & -0.14 \\ -0.02 & 0.86 & -0.04 & 0.06 \\ -0.12 & -0.04 & 0.72 & -0.08 \\ -0.14 & 0.06 & -0.08 & 0.74 \end{pmatrix}$$

and $b = (0.76; 0.08; 1.12; 0.68)$. Here we assume the tolerance $\varepsilon = 10^{-8}$.

Problem 6. Use *FR* Conjugate Gradient Method to minimize the function $f(x)$ which has three variables, that is, $x = (x_1; x_2; x_3)$. In the first iteration, we obtain $d_0 = (1; -1; 2)$ and then we obtain the new iteration point x^1 along the search direction d_0 with exact line

search. Let

$$\frac{\partial f(x^1)}{\partial x_1} = -2, \quad \frac{\partial f(x^1)}{\partial x_2} = -2.$$

Please write out the search direction d_1 at the point x^1 .

Problem 7. Let $Q \succ 0$ and the set of nonzero vectors d_1, \dots, d_n be Q -orthogonal. Prove that

$$Q^{-1} = \sum_{i=1}^n \frac{d_i d_i^T}{d_i^T Q d_i}.$$

Problem 8. We wish to find a hyper-plane $\omega^T x + \beta = 0$ to separate some data points $a_i \in R^d, i = 1, \dots, n$. If a clean separation is possible, we can formulate the problem as:

$$\begin{aligned} & \text{minimize}_{\omega, \beta} \quad \frac{1}{2} \|\omega\|^2 \\ & \text{subject to} \quad y_i(a_i^T \omega + \beta) \geq 1, \quad i = 1, \dots, n. \end{aligned}$$

Please write out its Lagrange dual problem.