EE8725 HW 5

Solve the following optimization problem using the interior point algorithm:

 $\min f(x_1, x_2) = 0.25x_1^2 + x_2^2$ $h(x_1, x_2) = x_2 - 0.05x_1^2 - 0.5x_1 + 2 \le 0$

See attached pages for theory of interior point algorithm

8.1 APPENDIX 8A: Interior Point Method

The interior point method converts the inequality constraints to equality constraints using the same technique of adding a slack variable to each constraint. Then a "penalty function" is added to the objective function so that the equations become:

Minimize: $f_{\mu} = f(\underline{x}) - \mu^k \sum_{i=1}^{N_{ineq}} \ln(s_i)$, where N_{ineq} is the number if inequality

constraints.

Subject to: $g(\underline{x}) = 0$

and

where $\mu^k \ge 0$ is the *barrier parameter* which is forced to decrease toward zero as the algorithm iterates to a solution (k is the iteration counter). The Lagrangian now looks like this:

$$L_{\mu} = f(x) - \mu^{k} \sum_{i=1}^{N_{ineq}} \ln(s_{i}) + \lambda^{T} g(x) + \gamma^{T} (h(x) + s)$$

And the Gradient of this Lagrangian becomes:

h(x) + s = 0

 $\nabla_{x}L_{\mu} = \nabla_{x}f(x) + \nabla_{x}g(x)\lambda + \nabla_{x}h(x)\gamma = 0$ $\nabla_{\lambda}L_{\mu} = g(x) = 0$ $\nabla_{\mu}L_{\mu} = h(x) + s = 0$ $\nabla_{s}L_{\mu} = -\mu^{k}S^{-1}e + \gamma = 0$ Where $S = \begin{bmatrix} s_{1} & & \\ & s_{2} & \\ & & \ddots \end{bmatrix}$ and $e = \begin{bmatrix} 1\\ 1\\ 1\\ \vdots \end{bmatrix}$

The algorithm at iteration k requires the solution of the following set of equations:

$$\begin{bmatrix} \nabla_x^2 L_{\mu} & \nabla_x g(x) & \nabla_x h(x) & 0 \\ \nabla_x g(x)^T & 0 & 0 & 0 \\ \nabla_x h(x)^T & 0 & 0 & I \\ 0 & 0 & I & \nabla_s^2 L_{\mu} \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta \lambda \\ \Delta \gamma \\ \Delta s \end{bmatrix} = \begin{bmatrix} \nabla_x L_{\mu} \\ \nabla_y L_{\mu} \\ \nabla_s L_{\mu} \end{bmatrix}$$

Solution of the Optimization Problem of Appendix 3A using Interior Point Algorithm¹

The optimization problem of Appendix 3A with one equality constraint and one inequality constraint is as given below:

 $\min f(x_1, x_2) = 0.25x_1^2 + x_2^2$ $g(x_1, x_2) = 5 - x_1 - x_2 = 0$ $h(x_1, x_2) = x_1 + 0.2x_2 - 3 \le 0$

To make this a more interesting problem we shall drop the equality constraint and reverse the direction of the inequality constraint, we then solve the problem:

 $\min f(x_1, x_2) = 0.25x_1^2 + x_2^2$ $h(x_1, x_2) = -x_1 - 0.2x_2 + 3 \le 0$

First we add a slack variable to the inequality constraint and the log barrier function to the objective:

$$\min f(x_1, x_2) = 0.25x_1^2 + x_2^2 - \mu^k \ln(s)$$

$$h(x_1, x_2) = -x_1 - 0.2x_2 + 3 + s = 0$$

The Lagrangian function is:

$$L(x_1, x_2, \gamma, s) = 0.25x_1^2 + x_2^2 - \mu^k \ln(s) + \gamma \left(-x_1 - 0.2x_2 + 3 + s\right)$$

And the minimum is found at the point $(x_1^*, x_2^*, \gamma^*, s^*)$ where:

$$\nabla_{x_1} L_{\mu} = 0.5 x_1^* - \gamma^* = 0$$

¹ This solution was done for a class homework problem by Volker Landenberger

$$\nabla_{x_2} L_{\mu} = 2x_2^* - 0.2\gamma^* = 0$$
$$\nabla_{\lambda} L_{\mu} = -x_1^* - 0.2x_2^* + 3 + s^* = 0$$
$$\nabla_{s} L_{\mu} = -\frac{\mu^k}{s^*} + \gamma^* = 0$$

Starting at $x_1 = 10$, $x_2 = 10$, and $\mu = 20$ this solution of these equations with a reduction of μ by one half each iteration converges to the solution:

 $\begin{aligned} x_1 &= 2.9704 \\ x_2 &= 0.1485 \\ \gamma &= 1.4852 \\ \mu &= 0.0002 \\ s &= 0.0001 \end{aligned}$