

## Homework 18 (Augmented Lagrangian)

Consider the problem:

$$\text{Minimize } f(\mathbf{x}) = \frac{1}{2} \left[ x_1^2 + \frac{1}{3} x_2^2 \right]$$

Subject to  $x_1 + x_2 = 1$

Use an unconstrained minimization method such as the conjugate gradient algorithm or the BFGS Q-N one to write a small code in order to compare method of augmented Lagrangian with that of quadratic penalty where the augmented Lagrangian is given by:

$$L(x, c_k, \lambda_k) = \frac{1}{2} \left[ x_1^2 + \frac{1}{3} x_2^2 \right] + \lambda_k (x_1 + x_2 - 1) + \frac{1}{2} c_k ((x_1 + x_2 - 1))^2$$

In the quadratic penalty method  $\lambda_k = 0$  for all k.

Use the following updating in the Augmented Lagrangian for the multipliers:

$$\lambda_{k+1} = \lambda_k + c_k (x_1^{(k)} + x_2^{(k)} - 1)$$

$$\lambda_0 = 0$$

For the penalty coefficients try 2 options:

$$c_{k+1} = 10c_k$$

$$c_0 = 0.2$$

$$c_k = 0.1 \times 2^k$$

Or try  $c_k = 0.1 \times 4^k$

$$c_k = 0.1 \times 8^k$$

Put the computational results in a form of table for k,  $x_1^{(k)}$ ,  $x_2^{(k)}$  for about 10-15 iterations for both methods and the different update variants.

Using first order conditions minimization of  $L(x, c_k, \lambda_k)$  yields:

$$x_1^{(k)} = \frac{c_k - \lambda_k}{1 + 4c_k} \quad x_2^{(k)} = \frac{3(c_k - \lambda_k)}{1 + 4c_k}$$

So it is evident that the optimal solution is:

$$x^* = (0.25, 0.75)$$

$$\lambda^* = -0.25$$

Your results should show (if your coding is correct) that method of augmented Lagrangian requires a smaller number of minimizations to obtain the solution and that the number of minimizations for both penalty and A-L methods decreases when the penalty parameter is increased at a faster rate.

However effects of ill-conditioning are felt more under these circumstances when the unconstrained minimization is carried out numerically.