

SSY281 - Model Predictive Control

Micro-Assignment MA06 - Optimization Basics and Convexity

Lucas Rath - **Group 09**

Question 1 For a generic unconstrained optimization problem, what is a necessary condition for x to be a solution?

For a generic unconstrained optimization problem, there exist a solution x^* if the problem is bounded below. That means that: $\inf f(x) > -\infty$

Question 2 Are the KKT conditions necessary and sufficient optimality conditions for any type of constrained optimization problem?

No, the KKT conditions are only necessary and sufficient for convex problems, since the local optimum will be also the global optimum.

Question 3 Write the Lagrangian for problem 40 in slide 69.

The Lagrangian is simply given by:

$$\mathcal{L}(x, \mu, \lambda) = f(x) + \mu^T g(x) + \lambda^T h(x) \quad (1)$$

Question 4 Write the KKT conditions and show that they hold at the solution you have found with the graphical method.

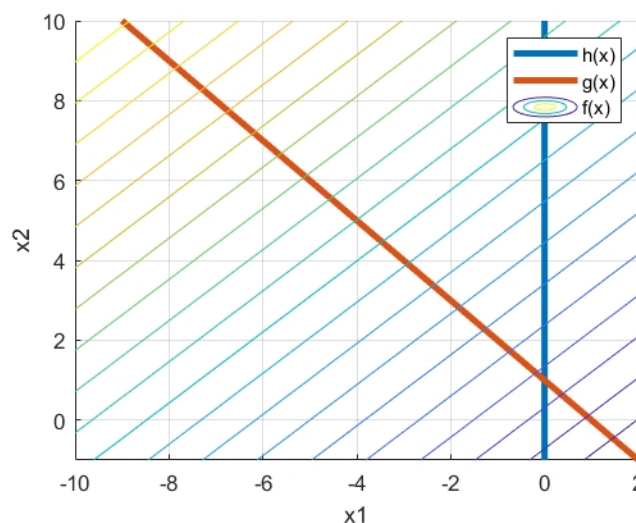


Figure 1: Solution of the constrained minimization problem

As can be verified in the picture, the optimal solution is given by $x^* = [0, 1]$.

The KKT conditions are given by:

$$\nabla f(x^*) + \nabla g(x^*)\mu^* + \nabla h(x^*)\lambda^* = 0 \quad (2)$$

$$\begin{bmatrix} 2 * x1^* \\ 2 * x2^* \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} \mu^* + \begin{bmatrix} 1 \\ 1 \end{bmatrix} \lambda^* = 0 \quad (3)$$

$$\begin{bmatrix} 0 \\ 2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} \mu^* + \begin{bmatrix} 1 \\ 1 \end{bmatrix} \lambda^* = 0 \quad (4)$$

$$\mu^* + \lambda^* = 0 \quad \text{and} \quad 2 + \lambda^* = 0 \quad (5)$$

What implies that $\lambda^* = -2$ and $\mu^* = 2$, which satisfies all the KKT conditions, in particular $\mu^* \geq 0$