## Optimization - EE5327

Srujana B - MA17BTECH11001 J. Sai Vyshnavi - MA17BTECH11005

20/02/2019

## Question 3.3

▶ Define L 
$$(\mathbf{x}, \lambda) = f(\mathbf{x}) - \lambda g(\mathbf{x})$$
 and  $\nabla = \begin{bmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \frac{\partial}{\partial \lambda} \end{bmatrix}$ .

Solve the equations  $\nabla L(\mathbf{x}, \lambda) = 0$ .

where 
$$f(\mathbf{x}) = (x_1 - 8)^2 + (x_2 - 6)^2$$
 and  $g(\mathbf{x}) = x_1 + x_2 - 9 = 0$ ,  $\mathbf{x} = (x_1 \ x_2)^T$ .

How is this related to problem 3.1? What is the sign of  $\lambda$ ? L is known as the Lagrangian and the above technique is known as the Method of Lagrange Multipliers,

L 
$$(\mathbf{x}, \lambda) = (x_1 - 8)^2 + (x_2 - 6)^2 - \lambda(x_1 + x_2 - 9)$$

$$\nabla L(\mathbf{x}, \lambda) = \begin{bmatrix} 2x_1 - 16 - \lambda \\ 2x_2 - 12 - \lambda \\ x_1 + x_2 - 9 \end{bmatrix}$$
Equating  $\nabla L(\mathbf{x}, \lambda)$  to 0

$$\begin{bmatrix} 2 & 0 & -1 \\ 0 & 2 & -1 \\ 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \lambda \end{bmatrix} = \begin{bmatrix} 16 \\ 12 \\ 9 \end{bmatrix}$$

This is of the form Ax = b

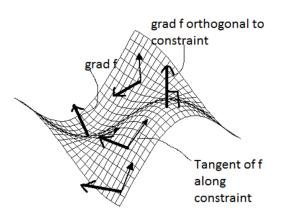
Solving the above equations, we get 
$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \lambda \end{bmatrix} = \begin{bmatrix} \frac{11}{2} \\ \frac{7}{2} \\ -5 \end{bmatrix}$$

$$\lambda = -5i.e.$$
 it is negative

- Let, maximize or minimize a function f(x,y) under a constraint path g(x,y)=c.
- Now, Gradient is the multidimensional rate of change of given function.

# **Figure**

(27).png



- ▶ So,  $\nabla f(x,y)$  is a vector that indicates direction to maximize f(x,y). But due to constraint g(x,y)=c allowed direction is to move along the tangents to constraint g(x,y)=c which is orthogonal to  $\nabla g(x,y)$ .
- ▶ Therefore, to maximize f(x,y), directional derivative along the path df must be zero, which means  $\nabla f(x,y,z)$  must be orthogonal to g(x,y,z)=c i.e. parallel to  $\nabla g(x,y,z)$ .

- Now, if two vectors are parallel they can be defined linearly(scalar multiply). Thus,  $\nabla f = \lambda \nabla g$  where  $\lambda$  is the Lagrange multiplier Lagrange function is defined by L(x,y, ) = f(x,y)  $\lambda g(x,y)$
- ▶ So,  $\nabla L = 0$