$$\left.\frac{\partial f^*(r)}{\partial r_j} = \left.\left|\frac{\partial f(x,r)}{\partial r_j}\right|_{x=x^*(r)}, where \ j=1,\dots,n\right.$$

This can prove that partial derivatives exist

## **Chapter 14 Optimization Constrained**

## Variable and Constrains

A general way of writing a problem would be

$$Max(min)f(x_1,...,x_n)$$
 when  $h(x_1,...,x_n) = a$ 

Using an example:

$$\mathcal{L}(x,y,z) = (x-1)^2 + (y-2)^2 + (z-3)^2 - \gamma(x^2 + y^2 + z)$$

Then first step is to find the first derivatives:

$$\mathcal{L}_{1}^{\prime}(x,y,z) = 2(x-1) - 2xy = 0$$

$$\mathcal{L}'_1(x,y,z) = 2(y-2) - 2yy = 0$$

$$\mathcal{L}_{1}^{I}(x, y, z) = 2(z-3) - \gamma = 0$$

$$x^2 + y^2 = -z$$

Now solve the following to get the values of  $x_t y_t z$  and y

In case of Constrains just subtract the sum of all the constrains from the equation and follow the same steps

## **What's Comparative Statics**

Just one important thing to remember:

$$\frac{\partial f^*(r)}{\partial r_j} = \left[\frac{\partial f(x,r)}{\partial r_j}\right]_{x = x^*(r)}, where \ j = 1, \dots, n$$

## **Non Linear Programming**

For  $\max f(x)$  subject to  $h_s(x) \le a$ , where  $s = 1, \dots, n$  and

x stands for  $(x_1, \dots, x_n)$ 

1. write the Lagrangian

$$\mathcal{L}(x) = f(x) - \sum_{s=1}^{n} \gamma_s (h_s(x) - \alpha_s)$$

2. Equate all the partial derivatives to 0

$$\frac{\partial \mathcal{L}\left(\mathbf{x}\right)}{\partial x_{j}} = \frac{\partial f\left(\mathbf{x}\right)}{\partial x_{j}} - \sum_{s=1}^{n} \gamma_{s} \frac{\partial h_{s}(\mathbf{x})}{\partial x_{j}} = \mathbf{0}, \quad \text{where j= 1,....,m}$$

3. Apply the complementary slackness conditions:

$$\gamma_s \ge 0$$
, if  $h_s(x) < a$  then  $\gamma_s = 0$ 

4. Satisfy the constrain

$$h_s(x) \le a_s$$

We should find all vectors x for which all the above are sure, then it could be considered as a possible solution.