

Lesson 11

Lagrange's Multiplier Rule / Constrained Optimization

11.1 Introduction

We present an introduction to optimization problems that involve finding a maximum or a minimum value of an objective function $f(x, y)$ subject to a constraint of the form $g(x, y) = k$.

Maximum and Minimum. Finding optimum values of the function $f(x, y)$

without a constraint is a well known problem in calculus. One would normally use the gradient to find critical points (gradient (∇f) vanishes). Then check all stationary and boundary points to find optimum values.

Example 1. $\nabla f = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right)$

$$f(x, y) = 2x^2 + y^2, f_x(x, y) = 4x = 0, f_y(x, y) = 2y = 0, f(x, y)$$

has a critical/ stationary point at $(0, 0)$.

The Hessian: A common method of determining whether or not a function has an extreme value at a stationary point is to evaluate the hessian of the function of n variables at that point. where the hessian is defined as

$$H(f) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} & \dots & \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_n^2} \end{pmatrix}$$

A square matrix of order $n \times n$ is said to be positive definite if its leading principal minors are all positive.

For $n=2$, we have

$$H(f) = \begin{pmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{pmatrix}$$

Second Derivative Test: The Second derivative test determines the optimality of stationary point \mathbf{x} according to the following rules:

Let $\frac{\partial^2 f}{\partial x^2} = A$, $\frac{\partial^2 f}{\partial x \partial y} = B$, $\frac{\partial^2 f}{\partial y^2} = C$, and $\frac{\partial f}{\partial x} = \frac{\partial f}{\partial y} = 0$ at the point (x, y) , then

1. If $A > 0$ and $AC - B^2 > 0$ at the point (x, y) , then f has a local minimum at (x, y) .
2. If $A < 0$ and $AC - B^2 > 0$ at the point (x, y) , then f has a local maximum at (x, y) .
3. If $AC - B^2 < 0$ at (x, y) , then (x, y) is a saddle point of f .
4. If $AC - B^2 = 0$, further investigation is required.

In the above Example 1,

$$\begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix}$$

Therefore $f(x, y)$ has a minimum at $(0, 0)$ as $4 > 0$ and determinant of the matrix is $8 > 0$.

11.1.1 Constrained Maximum and Minimum

When finding the extreme values of $f(x, y)$ subject to a constraint $g(x, y) = k$, the stationary points found above will not work. This new problem can be thought of as finding extreme values of $f(x, y)$ when the point (x, y) is restricted to lie on the

surface $g(x, y) = k$. The value of $f(x, y)$ is maximized (minimized) when the surfaces touch each other, i.e., they have a common tangent for line.

This means that the surfaces, gradient vectors at that point are parallel, hence,

$$\nabla f(x, y) = \lambda \nabla g(x, y)$$

The number λ in the equation is known as the Lagrange multiplier.

11.2 Lagrange multiplier method

The Lagrange multiplier method solves the constrained optimization problem by transforming it into a non-constrained optimization problem of the form:

$$L(x, y, \lambda) = f(x, y) + \lambda(k - g(x, y))$$

or $(g - k)$. Then finding the gradient and Hessian as was done above will determine any optimum values of $L(x, y, \lambda)$.

Suppose we want to find optimum values for the following:

Example 11.2: $f(x, y) = 2x^2 + y^2$ subject to $x + y = 1$.

Then the Lagrangian method will result in a non-constrained function.

$L(x, y, \lambda) = 2x^2 + y^2 + \lambda(1 - x - y)$. The gradient for this new function is

$$\frac{\partial L}{\partial x} = 4x - \lambda = 0$$

$$\frac{\partial L}{\partial y} = 2y - \lambda = 0$$

$$\frac{\partial L}{\partial \lambda} = 1 - x - y = 0$$

Solving x, y, λ , we obtain $x = \frac{1}{3}$, $y = \frac{2}{3}$ and $\lambda = \frac{4}{3}$.

The Hessian matrix at the stationary point

$$H(L) = \begin{pmatrix} 4 & 0 & -1 \\ 0 & 2 & -1 \\ -1 & -1 & 0 \end{pmatrix}$$

Since $H(L)$ is positive definite the solution $x = \frac{1}{3}$, $y = \frac{2}{3}$ minimizes

$f(x, y) = 2x^2 + y^2$ subject to $x + y = 1$ with $f\left(\frac{1}{3}, \frac{2}{3}\right) = \frac{2}{3}$

Example 11.3: Find the rectangle of parameter l which has maximum area i.e.,
Maximize xy subject to

$$2(x + y) = l$$

Solution:

$$L(x, y, \lambda) = xy + \lambda 2(x + y) - l).$$

$$\frac{\partial L}{\partial x} = y + 2\lambda = 0$$

$$\frac{\partial L}{\partial y} = x + 2\lambda = 0$$

$$\text{i.e., } x = y = -2\lambda \quad \text{i.e., } -8\lambda = l \Rightarrow \lambda = -\frac{l}{8}.$$

$x = y = \frac{l}{4}$, so that the rectangle of maximum area is a square.

Example 11.4 Find the shortest distance from the point (1,0) to the parabola $y^2 = 4x$, i.e., Minimize $(x - 1)^2 + y^2$ subject to $y^2 = 4x$.

$$L(x, y, \lambda) = (x - 1)^2 + y^2 + \lambda(y^2 - 4x)$$

$$\frac{\partial L}{\partial x} = 2(x - 1) - 4\lambda = 0, \quad \frac{\partial L}{\partial y} = 2y + 2\lambda y = 0$$

$$y^2 - 4x = 0$$

Now $2y + 2\lambda y = 0 \Rightarrow y = 0$ or $\lambda = -1$

If $\lambda = -1$, then $x = -1$, from $2(x - 1) - 4\lambda = 0$

Hence $y = 0 \Rightarrow x = 0$

$x = -1$

Now $y^2 = 4^n, y^2 = -4$, so $y = \sqrt{-4}$ not possible no real value.

Hence $y = 0, x = 0$,

i.e., $\lambda = -\frac{1}{2}$

Hence the only solution is $x = 0, y = 0, \lambda = -\frac{1}{2}$ and the required distance is unity.

Questions: Answer the following question

1. Determine the maximum value of the n -th root of a product of numbers

x_1, x_2, \dots, x_n provided that their sum is equal to a given number a . Thus the

problem is stated as follows: it is required to find the maximum of the function

$z = \sqrt[n]{x_1 \cdot x_2 \cdots x_n}$ subject to $\sum_{i=1}^n x_i - a = 0, x_i > 0$, for all i .

References

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Widder, D.V. (2002). *Advance Calculus* 2nd Edition, Publishers, PHI, India.

Piskunov, N. (1996). *Differential and Integral Calculus Vol I, & II*, Publishers, CBS, India.

Suggested Readings

Tom M. Apostol (2003). *Calculus, Volume II* Second Editions, Publishers, John Willey & Sons, Singapore.