## <span id="page-0-0"></span>The Method of Lagrange Multiplier

## Feng Li

Shandong University

fli@sdu.edu.cn

October 29, 2024

In mathematical optimization, the method of Lagrange multipliers is a strategy for finding the local maxima and minima of a function subject to equation constraints

max 
$$
f(x)
$$
  
s.t.  $g_i(x) = 0, i = 1, 2, \dots, m$ 

• It is named after the mathematician Joseph-Louis Lagrange

- The basic idea is to convert a constrained problem into a form such that the derivative test of an unconstrained problem can still be applied
- In general, the Lagrangian (or Lagrangian function) is defined by

$$
\mathcal{L}(x,\lambda)=f(x)+\sum_{i=1}^m\lambda_i g_i(x)
$$

where  $\lambda_i$  is called the Lagrange multiplier

 $\bullet$  In order to find the maximum (or minimum) of function  $f$  subject to the equality constraints  $g_i(x) = 0$ , find the stationary points of  $\mathcal L$ considered as a function of x and the Lagrange multiplier  $\lambda$ 

$$
\frac{\partial \mathcal{L}}{\partial x} = 0 \text{ and } \frac{\partial L}{\partial \lambda} = 0
$$

**•** Equivalently,

$$
\frac{\partial f(x)}{\partial x} + \sum_{i=1}^{m} \lambda_i \frac{\partial g_i(x)}{\partial x} = 0 \text{ and } g_i(x) = 0, \forall i = 1, 2, \cdots, m
$$

A rectangular box without a lid is to be made from  $27m^2$  of cardboard. Find the maximum volume of such a box

## Theorem (Lagrange Multiplier Theorem)

Let  $f: \mathbb{R}^n \to \mathbb{R}$  be the objective function,  $g: \mathbb{R}^n \to \mathbb{R}^m$  be the constraint function, both of which have continuous fist derivatives. Let  $x^*$  be an optimal solution to the following optimization problem

$$
\begin{array}{ll}\n\text{max} & f(x) \\
\text{s.t.} & g(x) = 0\n\end{array}
$$

such that  $\mathsf{Rank}(\mathsf{Dg}(x^*)) = m < n$  for the matrix of partial derivatives  $\left[\mathsf{D} g(x^*)\right]_{j,k} = \frac{\partial g_j}{\partial x_k}$  $\frac{\partial g_j}{\partial x_k}$ . Then, there exist a unique Lagrange multiplier  $\lambda^*\in\mathbb{R}^m$ such that

$$
Df(x^*) = (\lambda^*)^T Dg(x^*)
$$

- At any local maximum (or minimum) of the function evaluated under the equality constraints, if constraint qualification applies, then the gradient of the function (at that point) can be expressed as a linear combination of the gradients of the constraints (at that point), with the Lagrange multiplier serving as coefficients
- This is equivalent to saying that, any direction perpendicular to all gradients of the constraints is also perpendicular to the gradient of the function
- More details can be found at [https://en.wikipedia.org/wiki/](https://en.wikipedia.org/wiki/Lagrange_multiplier) [Lagrange\\_multiplier](https://en.wikipedia.org/wiki/Lagrange_multiplier)

## <span id="page-7-0"></span>Thanks!

Q & A