Lecture 9: Max-Min and Least Squares

1 Application of Partial Derivatives: Optimization Problem

1.1 Max-Min

Optimization problems refer to the tasks of finding the maximum or minimum point of a function. Here we focus on finding the maximum or minimum point of multivariable functions, such as f(x,y).

Another concept is local max/min point, which means the function value of this point is greater/less than the function values of adjacent points, but it is not necessarily the greatest or least function value in the function domain.

Theorem. If a point in a function f is a local max/min point, then all of the partial derivatives at this point are equal to 0.

Proof:

We will prove it by contradiction. Suppose there is a local max/min point in a function $f(x_1, x_2, ..., x_n)$, and one of its partial derivative $\frac{\partial f}{\partial x_i}$ is not 0. According to the definition of partial derivatives, in that direction the function values of the two adjacent point are

$$f(x_1, x_2, ..., x_i + \Delta x_i, ..., x_n) = f(x_1, x_2, ..., x_n) + \frac{\partial f}{\partial x_i} \Delta x_i$$

$$f(x_1, x_2, ..., x_i - \Delta x_i, ..., x_n) = f(x_1, x_2, ..., x_n) - \frac{\partial f}{\partial x_i} \Delta x_i$$

Therefore, one of them is less than the local max/min point and the other is greater than the local max/min point. Hence, it is not a local max/min point. There is a contradiction.

Therefore, it is proved that if a point in a function f is a local max/min point, then all of the partial derivatives at this point are equal to 0.

Therefore, A point is a local max/min point

- \Rightarrow All partial derivatives at that point are 0
- ⇔ The tangent plane at that point is horizontal
- \iff The point is a critical point.

1.2 Critical Points

Definition. A point is a critical point of a function f if and only if all the partial derivatives on this point are equal to 0.

For a multivariable function with two independent variables f(x,y), a point (x_0,y_0) is a critical point if $\frac{\partial f}{\partial x}(x_0,y_0)=0$ and $\frac{\partial f}{\partial y}(x_0,y_0)=0$.

Example 1. Find all the critical points of the function $f(x,y) = x^2 - 2xy + 3y^2 + 2x - 2y$. Solution:

$$\frac{\partial f}{\partial x} = 2x - 2y + 2$$
$$\frac{\partial f}{\partial y} = -2x + 6y - 2$$

For critical points, they should satisfy

$$\begin{cases} \frac{\partial f}{\partial x} = 2x - 2y + 2 = 0\\ \frac{\partial f}{\partial y} = -2x + 6y - 2 = 0 \end{cases}$$

Solving this linear system, the result is (-1,0). Therefore, there are only one critical point for this function, which is (-1,0).

1.3 Types of Critical Points

A critical point of a function f can be

- local maximum point
- local minimum point
- saddle point

Question 1. Is there any other possibilities for a critical point of a multivariable function? If not, how to prove it?

How to determine the type of a critical point?

- Test the second derivative (cover in the next lecture).
- Find the codomain of the function and compare the function value on the critical point with the codomain.

Example 2. Find the type of the critical point (-1,0) of the function $f(x,y) = x^2 - 2xy + 3y^2 + 2x - 2y$.

Solution:

The function value of the critical point (-1,0) is

$$f(-1,0) = (-1)^2 - 2 \times (-1) \times 0 + 3 \times 0^2 + 2 \times (-1) - 2 \times 0$$

= 1 - 2
= -1

To find the codomain of the function, we can transform its formula

$$f(x,y) = x^{2} - 2xy + 3y^{2} + 2x - 2y$$

$$= (x - y)^{2} + 2y^{2} + 2x - 2y$$

$$= ((x - y)^{2} + 2(x - y) + 1) + 2y^{2} - 1$$

$$= (x - y + 1)^{2} + 2y^{2} - 1$$

Therefore, it is apparent that the codomain of the function is $[-1, \infty]$. Hence, the critical point (-1,0) is the minimum point of the function.

2 Example of Optimization Problems: Least Squares

Given a series of discrete points (x_1, y_1) , (x_2, y_2) , ..., (x_n, y_n) , find the "best fit" line y = ax + b.

The task is to find the best values of a and b in the line equation y = ax + b. We can define a distance function about the line equation y = ax + b, and find proper values of a and b to minimize the distance function, in other words, find the minimum point (a_0, b_0) .

One way to define the distance function about the line equation y = ax + b is to use the total square deviation, where deviation means the vertical distance between an actual point and the line y = ax + b, described by the formula $y_i - (ax_i + b)$. This method is called **least squares**.

Therefore, the distance function is

$$D = (y_1 - ax_1 - b)^2 + (y_2 - ax_2 - b)^2 + \dots + (y_n - ax_n - b)^2$$

$$= \sum_{i=1}^n (y_i - ax_i - b)^2$$

$$\frac{\partial D}{\partial a} = \sum_{i=1}^n -2x_i(y_i - ax_i - b)$$

$$\frac{\partial D}{\partial b} = \sum_{i=1}^n -2(y_i - ax_i - b)$$

To find the minimum point, we need to solve the following system of equations:

$$\begin{cases} \frac{\partial D}{\partial a} = \sum_{i=1}^{n} -2x_i(y_i - ax_i - b) = 0\\ \frac{\partial D}{\partial b} = \sum_{i=1}^{n} -2(y_i - ax_i - b) = 0 \end{cases}$$

which is a linear system of a and b. After simplified, it becomes

$$\begin{cases} (\sum_{i=1}^{n} x_i^2) a + (\sum_{i=1}^{n} x_i) b = \sum_{i=1}^{n} x_i y_i \\ (\sum_{i=1}^{n} x_i) a + nb = \sum_{i=1}^{n} y_i \end{cases}$$

Therefore, we can solve the linear system to get the critical points of the distance function, and hence get the minimum point (a_0, b_0) .

The least squares method can be more general. First of all, it can be used to approximate any polynomial functions, since the deviation is a linear function of unknown parameters. For example,

$$y = ax^{2} + bx + c$$
$$d_{i} = y_{i} - ax_{i}^{2} - bx_{i} - c$$

Second, for other functions, we can try to transform it into polynomial functions. A typical example is exponential functions:

$$y = ce^{ax}$$

$$\ln y = ax + \ln c$$

$$d_i = \ln y_i - ax_i - \ln c$$