

Constrained max/min

We should all recall the proposition:

Proposition. *Let D be a compact region in \mathbb{R}^n , with interior U and $f : D \rightarrow \mathbb{R}$ a function that is C^1 in U . Then the maximum and minimum values of f on D exist. Each occurs at either a critical point for f (where $\nabla f = \mathbf{0}$) in U or a point of ∂D .*

The result for constrained max/min is somewhat similar. Unfortunately, we must bring in the Implicit Function Theorem to explain it properly. Let $Y \subset \mathbb{R}^n$ be the intersection of level sets of m C^1 functions $g_1 = g_2 = \dots = g_m = 0$ (constraint conditions). Let f be a function that is C^1 in some neighborhood of Y . If these g_i 's satisfy a rank condition at $\mathbf{x} \in Y$, then in a neighborhood of x in \mathbb{R}^n , Y is equal to the graph of m coordinates of \mathbb{R}^n as a C^1 function of the other $n - m$. When $m = 1$, the "rank condition" is simply $\nabla g_1(\mathbf{x}) \neq \mathbf{0}$; in general, it is the assertion that the gradient vectors $\nabla g_i(\mathbf{x})$ point in independent directions.

We now give the constrained version of the proposition above in the case of $m = 1$:

Proposition. *Let $Y \subset \mathbb{R}^n$ be a compact level set of the C^1 function g , and f be a C^1 function that is defined in a neighborhood of Y . Then the maximum and minimum values of f on Y exist. They occur points where ∇f is a multiple of ∇g , or where $\nabla g = \mathbf{0}$.*

Proof. Suppose that $\mathbf{a} \in Y$, and that $\nabla g \neq \mathbf{0}$ at (and hence near) \mathbf{a} . Using the Implicit Function Theorem, we can write one of the variables, say x_n , as $h(x_1, \dots, x_{n-1})$ on Y (near \mathbf{a}). Thus, we are seeking critical points for a function of $n - 1$ variables, namely

$$F(x_1, \dots, x_{n-1}) = f(x_1, \dots, x_{n-1}, h(x_1, \dots, x_{n-1})),$$

with $G(x_1, \dots, x_{n-1}) = g(x_1, \dots, x_{n-1}, h(x_1, \dots, x_{n-1})) = 0$ determining the implicit differentiation of h . To be specific, writing F_i for the partial derivative F_{x_i} , etc., we have for $i < n$ by the Chain Rule

$$\begin{aligned} (*) \quad F_i(x_1, \dots, x_{n-1}) &= f_i(x_1, \dots, x_{n-1}, h(x_1, \dots, x_{n-1})) \\ &\quad + f_n(x_1, \dots, x_{n-1}, h(x_1, \dots, x_{n-1}))h_i(x_1, \dots, x_{n-1}) \end{aligned}$$

and

$$\begin{aligned} (**) \quad G_i(x_1, \dots, x_{n-1}) &= g_i(x_1, \dots, x_{n-1}, h(x_1, \dots, x_{n-1})) \\ &\quad + g_n(x_1, \dots, x_{n-1}, h(x_1, \dots, x_{n-1}))h_i(x_1, \dots, x_{n-1}) = 0 \end{aligned}$$

Where $\nabla F = \mathbf{0}$ (*) and (**) have the same format. We write in condensed format: $f_i + f_n h_i = 0$ and $g_i + g_n h_i = 0$, and $g_n \neq 0$. This gives $f_i/f_n = g_i/g_n$ (when defined). Take it from here.

In the example with $f(x, y) = y - x^2$ and $g(x, y) = x^2 + y^2 - 1$, we are led to solve $\nabla(f - \lambda g) = 0$. Here, that's

$$-2x - \lambda(2x) = 0 \quad \text{and} \quad 1 - \lambda(2y) = 0,$$

which yields $x = 0$ or $\lambda = -1$, and $y = 1/2\lambda$; and, of course, $x^2 + y^2 = 1$. The solutions are $(x, y) = (0, \pm 1)$ and $(x, y) = (\pm\sqrt{3}/2, -1/2)$. The values of $f(x, y)$ at these points are ± 1 , and $-5/4$. We can see the absolute maximum and minimum values of f on the unit circle.