

Unconstrained Optimization

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1 Introduction

While optimization is a broad topic, our goal in this lecture will be focused on a narrow issue. We will dwell in the realm of nicely-behaved (continuous and differentiable) functions. We will also restrict ourselves to think about local optima. We will also use compact notations for the partial derivatives.

$$f_x = \frac{\partial f}{\partial x} \quad f_y = \frac{\partial f}{\partial y} \quad f_{xx} = \frac{\partial^2 f}{\partial x^2} \quad f_{yy} = \frac{\partial^2 f}{\partial y^2} \quad f_{xy} = \frac{\partial^2 f}{\partial y \partial x} \quad f_{yx} = \frac{\partial^2 f}{\partial x \partial y}$$

2 Local Optima

Let (x_0, y_0) be a **stationary point** of the function f of two variables. Then, we should have (at this point):

Step 1: Find and set the first-order partial derivatives to zero to find (x_0, y_0) . Therefore, the necessary condition is:

$$\begin{aligned} f_x &= 0 \\ f_y &= 0 \end{aligned}$$

Step 2: In order to classify (or label) the stationary, we need all the second-order derivatives. We also need the *Hessian* of the function.

Let f be a twice-differentiable function of two variables. The **Hessian** of f at \mathbf{x} is

$$H(\mathbf{x}) = \begin{pmatrix} f_{xx}(\mathbf{x}) & f_{xy}(\mathbf{x}) \\ f_{yx}(\mathbf{x}) & f_{yy}(\mathbf{x}) \end{pmatrix}$$

We define the determinant of the Hessian as: $D = (f_{xx}f_{yy}) - (f_{xy})^2$.

Case	Second Order Conditions	Conclusion	Concavity
1.	$D > 0$ and $f_{xx} < 0$	Local Maximum	Concave.
2.	$D > 0$ and $f_{xx} > 0$	Local Minimum	Convex.
3.	$D < 0$	Saddle Point	Convex (one side), Concave (other side).
4.	$D = 0$	Inconclusive	-

3 Examples

1- $f(x, y) = (x - 1)^2 + (y - 1)^2$

To find the stationary points, we first compute the first-order partial derivatives and set them equal to zero.

$$\begin{aligned}
 f_x &= \frac{\partial}{\partial x}((x - 1)^2 + (y - 1)^2) \\
 &= 2(x - 1) \\
 f_y &= \frac{\partial}{\partial y}((x - 1)^2 + (y - 1)^2) \\
 &= 2(y - 1)
 \end{aligned}$$

We look for potential stationary points by setting both partial derivatives to zero.

$$\begin{aligned}
 f_x &= 0 \\
 \implies 2(x - 1) &= 0 \\
 \implies x &= 1 \\
 f_y &= 0 \\
 \implies 2(y - 1) &= 0 \\
 \implies y &= 1
 \end{aligned}$$

We got only one candidate point: $(1, 1)$.

Let's now test for the second order condition for which we need the second order partial derivatives.

$$\begin{array}{cccc}
 f_{xx} = \frac{\partial}{\partial x}(2(x - 1)) & f_{yy} = \frac{\partial}{\partial y}(2(y - 1)) & f_{xy} = \frac{\partial}{\partial y}(2(x - 1)) & f_{yx} = \frac{\partial}{\partial x}(2(y - 1)) \\
 = 2 & = 2 & = 0 & = 0
 \end{array}$$

We can see that $f_{xx} > 0$ and $f_{yy} > 0$. Therefore, it seems like $(1, 1)$ represents the local minimum, but we must also check the value of the determinant of the Hessian. $D > 0$ implies that this is indeed the **local minimum**.

$$2- f(x, y) = 60x + 34y - 6x^2 - 3y^2 - 4xy + 5.$$

First Order Partial Derivatives

$$f_x = 60 - 12x - 4y$$

$$f_y = 34 - 6y - 4x$$

Set these two partial derivatives to zero.

$$12x + 4y = 60$$

$$4x + 6y = 34$$

Solving these equations, we get: $x = 3, y = 4$.

We have only candidate stationary point: $(3, 4)$.

Second Order Partial Derivatives

$f_{xx} = \frac{\partial}{\partial x}(60 - 12x - 4y)$	$f_{yy} = \frac{\partial}{\partial y}(34 - 6y - 4x)$	$f_{xy} = \frac{\partial}{\partial y}(60 - 12x - 4y)$	$f_{yx} = \frac{\partial}{\partial x}(34 - 6y - 4x)$
$= -12$	$= -6$	$= -4$	$= -6$

$f_{xx} < 0, f_{yy} < 0$, and $D > 0 \implies$ **Local maximum**

$$3- f(x, y) = 4x^2 - xy + y^2 - x^3.$$

First Order Partial Derivatives

$$f_x = 8x - y - 3x^2$$

$$f_y = -x + 2y$$

Set these two partial derivatives to zero.

$$8x^2 - y - 3x^2 = 0$$

$$x = 2y$$

$$\implies 16y - y - 3(2y)^2 = 0$$

$$\implies 15y - 12y^2 = 0$$

$$\implies 3y(5 - 4y) = 0$$

$$\implies y = 0, \frac{5}{4}$$

$$\implies x = 0, \frac{5}{2}$$

We have two stationary points: $(0, 0)$ and $(\frac{5}{2}, \frac{5}{4})$.

Second Order Partial Derivatives

$f_{xx} = \frac{\partial}{\partial x}(8x - y - 3x^2)$	$f_{yy} = \frac{\partial}{\partial y}(-x + 2y)$	$f_{xy} = \frac{\partial}{\partial y}(8x - y - 3x^2)$	$f_{yx} = \frac{\partial}{\partial x}(-x + 2y)$
$= 8 - 6x$	$= 2$	$= -1$	$= -1$

Based on the values, we have a **local minimum** at $(0, 0)$ and a **saddle point** at $(\frac{5}{2}, \frac{5}{4})$.