

## Day 24

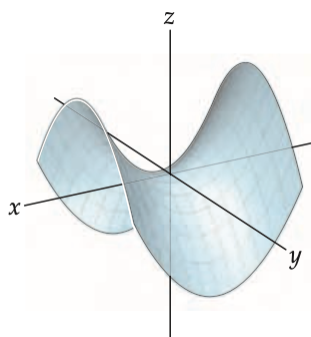
### 1. CLASSIFYING CRITICAL POINTS

Last time we saw the definition of the local extrema (i.e. local minima and maxima) of a function. We saw that local extrema only occur at critical points. If we've found a critical point, how can we tell whether it's a local minimum or local maximum (or both or neither)? The *second derivative test* will provide a criterion that answers this question in most cases. We will only state the second derivative test for functions of two variables. (There is a second derivative test for functions with three or more variables, but it requires some knowledge of linear algebra.) First we need a few more definitions:

#### Saddle Points of a Function of Two Variables

A point  $(x_0, y_0)$  in the domain of a function  $f(x, y)$  is called a **saddle point** of  $f$  if  $(x_0, y_0)$  is a stationary point of  $f$  at which there is neither a maximum nor a minimum.

The prototypical example of a function with a saddle point is the hyperbolic paraboloid  $f(x, y) = x^2 - y^2$ , whose graph looks like this:



This function has a saddle point at the origin  $(0, 0)$ . There the tangent plane is horizontal (i.e. the gradient is zero), but the function has neither a local maximum nor a local minimum.

Recall that a function of  $x$  and  $y$  has four second partial derivatives:  $f_{xx}$ ,  $f_{xy}$ ,  $f_{yx}$ , and  $f_{yy}$ . The second derivative test will involve all of these. It's convenient to organize these derivatives into a  $2 \times 2$  matrix, known as the Hessian matrix.

### The Hessian and the Discriminant of a Function of Two Variables

Let  $f(x, y)$  be a function with continuous second-order partial derivatives on some open set  $S$ .

(a) The **Hessian** of  $f$  is the  $2 \times 2$  matrix of second-order partial derivatives:

$$H_f = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial y \partial x} \\ \frac{\partial^2 f}{\partial x \partial y} & \frac{\partial^2 f}{\partial y^2} \end{bmatrix}.$$

(b) The **discriminant** of  $f$  is the determinant of the Hessian. That is,

$$\det(H_f) = \frac{\partial^2 f}{\partial x^2} \frac{\partial^2 f}{\partial y^2} - \left( \frac{\partial^2 f}{\partial x \partial y} \right)^2.$$

**In-class exercise<sup>1</sup>:** Let  $f(x, y) = x^2 + xy - y^2$

- Find the critical point(s) of  $f$ .
- At each critical point, find the Hessian matrix  $H_f$  and the discriminant  $\det(H_f)$ .

### THEOREM 12.45

#### The Second-Order Partial-Derivative Test for Classifying Stationary Points

Let  $f(x, y)$  be a function with continuous second-order partial derivatives on some open disk containing the point at  $(x_0, y_0)$ . If  $f$  has a stationary point at  $(x_0, y_0)$ , then

- $f$  has a relative maximum at  $(x_0, y_0)$  if  $\det(H_f(x_0, y_0)) > 0$  with  $f_{xx}(x_0, y_0) < 0$  or  $f_{yy}(x_0, y_0) < 0$ .
- $f$  has a relative minimum at  $(x_0, y_0)$  if  $\det(H_f(x_0, y_0)) > 0$  with  $f_{xx}(x_0, y_0) > 0$  or  $f_{yy}(x_0, y_0) > 0$ .
- $f$  has a saddle point at  $(x_0, y_0)$  if  $\det(H_f(x_0, y_0)) < 0$ .
- If  $\det(H_f(x_0, y_0)) = 0$ , no conclusion may be drawn about the behavior of  $f$  at  $(x_0, y_0)$ .

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<sup>1</sup> (a)  $(0, 0)$   
(b)  $-5$