

## Summary: Critical points

### Big Picture

When we are interested in finding the maximum or minimum value of a function, we look to see mathematically what properties of that function are necessary to have a maximum or minimum.

Local maxima and minima of a function  $f(x, y)$  occur at points where the gradient is zero (or undefined). We call points where the gradient is zero **critical points**. A critical point can be a local maximum, a local minimum, or neither, which is called a saddle point.

The gradient or level curves of a function give us graphical information about the behavior of a function that allows us to determine the type of critical point we have.

### Mechanics

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### Critical points

**Definition.** Let  $f(x, y)$  be a function of two variables. A **critical point** of  $f(x, y)$  is a point  $(x_0, y_0)$  at which  $\nabla f(x_0, y_0) = \vec{0}$ . In other words, when  $f_x(x_0, y_0) = 0$  and  $f_y(x_0, y_0) = 0$  simultaneously.

#### Extension to higher dimension: Critical points

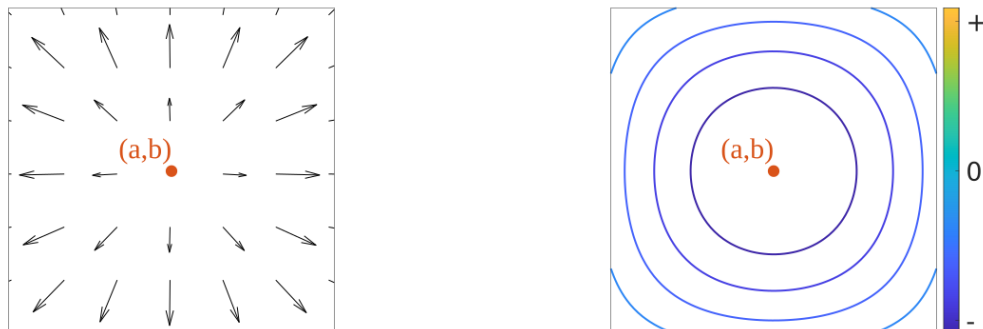
Let  $f(x_1, x_2, \dots, x_n)$  be a function of  $n$  variables. A **critical point** of  $f(x_1, x_2, \dots, x_n)$  is a point  $(x_1^*, x_2^*, \dots, x_n^*)$  at which  $\nabla f(x_1^*, x_2^*, \dots, x_n^*) = \vec{0}$ . In other words, when  $f_{x_1}(x_1^*, x_2^*, \dots, x_n^*) = 0$ ,  $f_{x_2}(x_1^*, x_2^*, \dots, x_n^*) = 0$ ,  $\dots$ , and  $f_{x_n}(x_1^*, x_2^*, \dots, x_n^*) = 0$  simultaneously.

### Graphical methods

Suppose  $(x, y) = (a, b)$  is a critical point of  $f(x, y)$  (meaning  $\nabla f(a, b) = \langle 0, 0 \rangle$ ).

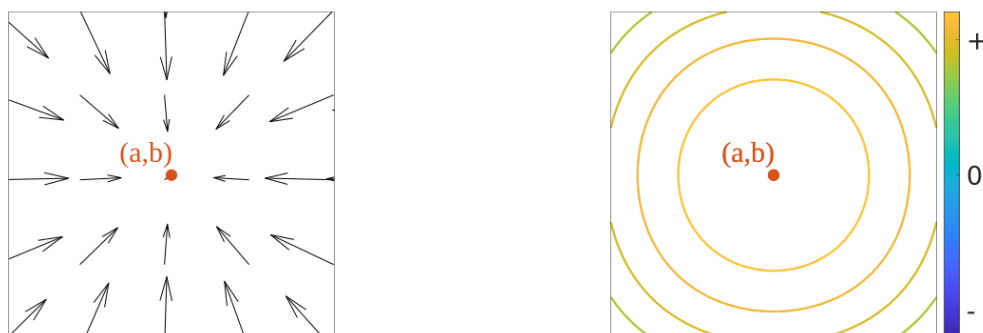
**Case 1:** If the vectors representing  $\nabla f(x, y)$  surrounding  $(a, b)$  are pointing away from  $(a, b)$ , then  $f(x, y)$  is decreasing as we approach  $(a, b)$  from every direction. This means  $(a, b)$  is a local minimum of  $f(x, y)$ .

The figure below on the left shows the gradient field near a local minimum. The figure below on the right shows the corresponding level curves.



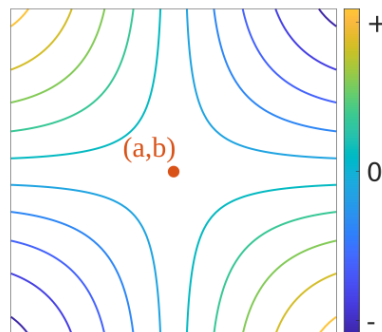
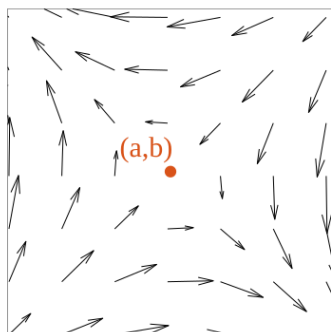
**Case 2:** If the vectors representing  $\nabla f(x, y)$  surrounding  $(a, b)$  are pointing towards  $(a, b)$ , then  $f(x, y)$  is increasing as we approach  $(a, b)$  from every direction. This means  $(a, b)$  is a local maximum of  $f(x, y)$ .

The figure below on the left shows the gradient field near a local maximum. The figure below on the right shows the corresponding level curves.



**Case 3:** If some vectors representing  $\nabla f(x, y)$  near  $(a, b)$  point towards  $(a, b)$  and some point away from  $(a, b)$ , then  $f(x, y)$  is increasing as we approach  $(a, b)$  from some directions and decreasing as we approach  $(a, b)$  from other directions. This means  $(a, b)$  is a saddle point of  $f(x, y)$ .

The figure below on the left shows the gradient field near a saddle point. The figure below on the right shows the corresponding level curves.



Ask yourself

## Ask Yourself

What does "finding critical points" have to do with "maximizing a function of two variables"?

What does it mean to "maximize a function of two variables"?