

10.6 Directional Derivatives and Gradients

In this section, we will introduce the concepts of the directional derivative and gradient.

The Directional Derivative

Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a function. The **directional derivative of f at the point \mathbf{x} in the direction of the unit vector $\hat{\mathbf{u}}$** denoted by $D_{\hat{\mathbf{u}}}f(\mathbf{x})$ and is defined as

$$D_{\hat{\mathbf{u}}}f(\mathbf{x}) = \lim_{h \rightarrow 0} \frac{f(\mathbf{x} + h\hat{\mathbf{u}}) - f(\mathbf{x})}{h}.$$

In particular, if $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is differentiable at the point $\mathbf{x} = (x_1, \dots, x_n)$ we have that

$$D_{\hat{\mathbf{u}}}f(\mathbf{x}) = \frac{\partial f}{\partial x_1}u_1 + \frac{\partial f}{\partial x_2}u_2 + \cdots + \frac{\partial f}{\partial x_n}u_n.$$

The Gradient

Suppose $f(\mathbf{x})$ is differentiable at the point $\mathbf{p} = (p_1, \dots, p_n)$. The **gradient of $f(\mathbf{x})$ at the point \mathbf{p}** is the vector $\nabla f(\mathbf{p})$ defined as

$$\nabla f(\mathbf{p}) = \frac{\partial f}{\partial x_1}\hat{\mathbf{x}}_1 + \cdots + \frac{\partial f}{\partial x_n}\hat{\mathbf{x}}_n.$$

In particular, the gradient vector points in the direction of greatest positive change in the function value.

Theorem: Gradient Adjunction Formula

If $f(\mathbf{x})$ is differentiable at a point $\mathbf{p} = (p_1, \dots, p_n)$ and $\hat{\mathbf{u}} = (u_1, \dots, u_n)$ is any unit vector, then we have that

$$D_{\hat{\mathbf{u}}}f(\mathbf{p}) = \nabla f(\mathbf{p}) \cdot \hat{\mathbf{u}} = \|\nabla f(\mathbf{p})\| \cos \theta,$$

where θ is the angle between the vectors $\nabla f(\mathbf{p})$ and $\hat{\mathbf{u}}$.

Question 1. For each of the following, find the directional derivative **using only the limit definition**.

(a) $f(x, y) = 5 - 2x^2 - \frac{1}{2}y^2$ at the point $(3, 4)$ in the direction of the unit vector $(\sqrt{2}/2, \sqrt{2}/2)$.

(b) $h(x, y) = e^x \sin(y)$ at the point $(1, \pi/2)$ in the direction of the unit vector $\hat{\mathbf{u}} = -\hat{\mathbf{x}}$.

(c) $f(x, y) = y^2 - x^3 - x^2$ at the point $(0, 0)$ in the direction of the unit vector $(\sqrt{2}/2, -\sqrt{2}/2)$.

(d) $f(x, y) = y^2 - x^3 - x^2$ at the point $(-1, 0)$ in the direction of the unit vector $(0, 1)$.

Question 2. Let $h(x, y) = x^2 + y^2 - 1$.

- (a) Sketch a contour plot of $h(x, y)$. Make sure to include contours corresponding to at least $h = -1$, $h = 0$, $h = 1$, $h = 2$, and $h = 3$.
- (b) Find the gradient of $h(x, y)$.
- (c) Find the gradient of $h(x, y)$ at the each points $(\pm 1, 0)$, $(0, \pm 1)$. Plot these vectors on your contour plot.
- (d) What do you notice about the magnitude of $\nabla h(x, y)$ at these points? The direction in which the gradient vectors point?

Question 3. For each of the following, compute the gradient.

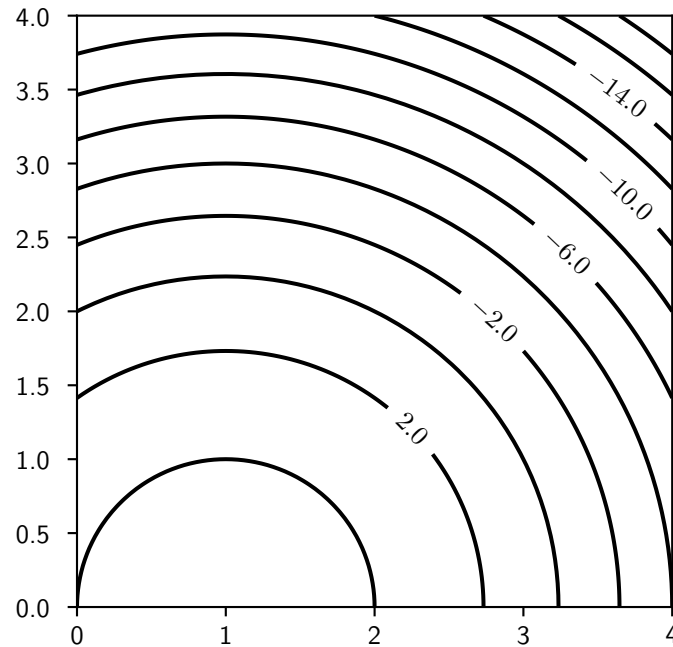
(a) $f(x, y) = x^2 + y^2$

(b) $g(x, y, z) = \sqrt{x^2 + y^2 + z^2}$

(c) $h(x, y) = \sin^2(\pi x) + \sin^2(\pi y)$.

(d) $u(x, y) = x^2 + 6xy - y^2$

Question 4. (Application:) In this problem, we will explore how we can use the gradient to maximize a function. For this example, let $f(x, y) = 4 - x^2 - y^2 + 2x$.



A contour plot of $f(x, y) = 4 - x^2 - y^2 + 2x$

- (a) Find the gradient of $f(x, y)$.
- (b) The first step in the algorithm is to pick an arbitrary point $\mathbf{p}_0 = (x_0, y_0)$ in the domain of our function. Let's pick the point $\mathbf{p}_0 = (2, 1)$ as our initial guess. Compute $\nabla f(\mathbf{p}_0)$. Plot the gradient vector at \mathbf{p}_0 on the contour plot.

- (c) We will now “update” our guess by moving a small distance h along the gradient vector and then see where we land. For this example, we’ll work with a fairly large step size, say $h = 0.25$. Let $\mathbf{p}_1 = \mathbf{p}_0 + h\nabla f(\mathbf{p}_0)$ be the next point. Compute \mathbf{p}_1 and $\nabla f(\mathbf{p}_1)$. Then plot $\nabla f(\mathbf{p}_1)$ on the contour plot.
- (d) We can continue this process. The general process we follow is that $\mathbf{p}_{n+1} = \mathbf{p}_n + h\nabla f(\mathbf{p}_n)$. Using this method, compute \mathbf{p}_n and $\nabla f(\mathbf{p}_n)$ for $n = 3, 4, 5$, and plot each of the gradient vectors on the contour plot.
- (e) This process will either continue until we reach a point where the gradient vector is zero, or will run off to infinity. Why would the process stop if we had $\nabla f(\mathbf{p}) = \mathbf{0}$ for some \mathbf{p} ?

10.7 Optimization

In this section, we will examine critical points of multivariable functions and how to find them.

Definition: Critical Points of Multivariable Functions

Let $f(x, y)$ be a differentiable function. A point $\mathbf{p} = (x_0, y_0)$ is a **critical point** of $f(\mathbf{x})$ if $\nabla f(\mathbf{p}) = \mathbf{0}$.
Let $\mathbf{p} = (x_0, y_0)$ be a critical point of $f(x, y)$.

- If $f(\mathbf{p}) \geq f(\mathbf{x})$ for all points \mathbf{x} near \mathbf{p} , then \mathbf{p} is a **local maximum** of f .
- If $f(\mathbf{p}) \leq f(\mathbf{x})$ for all points \mathbf{x} near \mathbf{p} , then \mathbf{p} is a **local minimum** of f .
- The critical point \mathbf{p} is a **saddle point** of $f(x, y)$ if it is neither a local maximum nor a local minimum.

Theorem: The Multivariable Second Derivative Test

Suppose we have a function $f(x, y)$ that is differentiable everywhere. Let $\mathbf{p} = (x_0, y_0)$ be a critical point of $f(x, y)$. Then, let $D = f_{xx}(\mathbf{p})f_{yy}(\mathbf{p}) - f_{xy}(\mathbf{p})^2$. (This D is called the *discriminant* or *Hessian* of $f(x, y)$ at \mathbf{p} .)

Then:

- If $D > 0$ and $f_{xx}(\mathbf{p}) < 0$ or $f_{yy}(\mathbf{p}) < 0$, then f has a local maximum at \mathbf{p}
- If $D > 0$ and $f_{xx}(\mathbf{p}) > 0$ or $f_{yy}(\mathbf{p}) > 0$, then f has a local minimum at \mathbf{p}
- If $D < 0$ then f has a saddle point at \mathbf{p}
- If $D = 0$ then this test yields no information about what happens at (x_0, y_0) .

Theorem: The Extreme Value Theorem

If $f(x, y)$ is continuous on a closed, bounded region $\Omega \subset \mathbb{R}^2$, then $f(x, y)$ attains its global maximum and minimum inside Ω .

Question 1. Find and classify all critical points of $f(x, y) = x^2y^2 - 6x^2y - 4xy^2 + 24xy$.

Question 2. Let a and b be any two real numbers, and let $f(x, y) = xy - ax - by$.

(a) How many local minimum points must $f(x, y)$ have?

(b) How many local maximum points must $f(x, y)$ have?

(c) How many saddle points must $f(x, y)$ have?

Question 3. Find and classify all critical points of the function $f(x, y) = 22xye^{-x^2-y^2}$ using the second derivative test.

Question 4. A twice-differentiable function $h(x, y)$ is called **harmonic** on a closed, bounded region Ω if $f_{xx}(x, y) + f_{yy}(x, y) = 0$ for each point (x, y) in Ω . Harmonic functions satisfy the extrema principle— all of their minima and maxima lie on the boundary of Ω and the only critical points h can have on the interior of Ω are the saddle points.

(a) Use the Extreme Value Theorem to show that $h(x, y)$ must have maximum **and** minimum values on the boundary of Ω .

(b) Use the extrema principle to argue that, if $h(x, y)$ is constant on the boundary of Ω , then $h(x, y)$ must be constant on all of Ω .

