

## Directional derivatives and gradient vectors (Sect. 14.5).

- ▶ Directional derivative of functions of two variables.
- ▶ Partial derivatives and directional derivatives.
- ▶ Directional derivative of functions of three variables.
- ▶ The gradient vector and directional derivatives.
- ▶ Properties of the the gradient vector.

## Directional derivative of functions of two variables.

**Remark:** The directional derivative generalizes the partial derivatives to any direction.

### Definition

The *directional derivative* of the function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  at the point  $P_0 = (x_0, y_0) \in D$  in the direction of a unit vector  $\mathbf{u} = \langle u_x, u_y \rangle$  is given by

$$(D_{\mathbf{u}}f)_{P_0} = \lim_{t \rightarrow 0} \frac{1}{t} [f(x_0 + u_x t, y_0 + u_y t) - f(x_0, y_0)],$$

if the limit exists.

**Notation:** The directional derivative is also denoted as

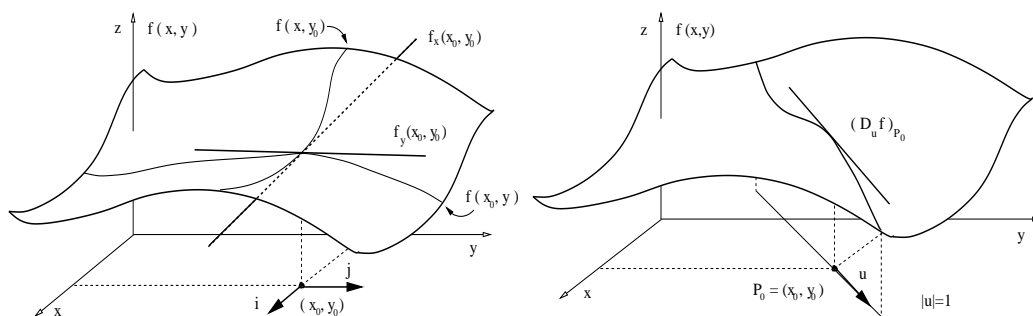
$$\left( \frac{df}{dt} \right)_{\mathbf{u}, P_0}.$$

## Directional derivatives generalize partial derivatives.

### Example

The partial derivatives  $f_x$  and  $f_y$  are particular cases of directional derivatives  $(D_{\mathbf{u}}f)_{P_0} = \lim_{t \rightarrow 0} \frac{1}{t} [f(x_0 + u_x t, y_0 + u_y t) - f(x_0, y_0)]$ :

- ▶  $\mathbf{u} = \langle 1, 0 \rangle = \mathbf{i}$ , then  $(D_{\mathbf{i}}f)_{P_0} = f_x(x_0, y_0)$ .
- ▶  $\mathbf{u} = \langle 0, 1 \rangle = \mathbf{j}$ , then  $(D_{\mathbf{j}}f)_{P_0} = f_y(x_0, y_0)$ .



## Directional derivative of functions of two variables.

**Remark:** The condition  $|\mathbf{u}| = 1$  implies that the parameter  $t$  in the line  $\mathbf{r}(t) = \langle x_0, y_0 \rangle + \mathbf{u}t$  is the distance between the points  $(x(t), y(t)) = (x_0 + u_x t, y_0 + u_y t)$  and  $(x_0, y_0)$ .

**Proof.**

$$d = |\langle x - x_0, y - y_0 \rangle| = |\langle u_x t, u_y t \rangle| = |t| |\mathbf{u}|,$$

that is,  $d = |t|$ . □

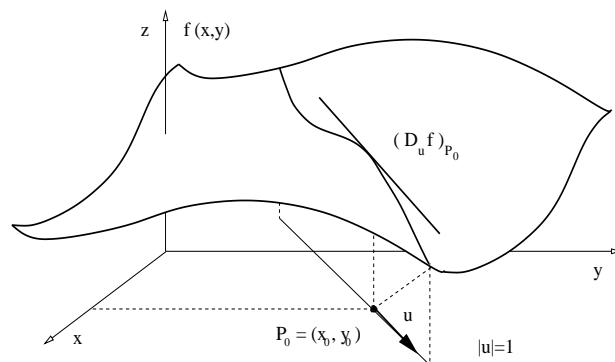
**Remark:** The directional derivative of  $f(x, y)$  at  $P_0 = (x_0, y_0)$  along  $\mathbf{u}$ , denoted as  $(D_{\mathbf{u}}f)_{P_0}$ , is the pointwise rate of change of  $f$  with respect to the **distance** along the line parallel to  $\mathbf{u}$  passing through  $(x_0, y_0)$ .

## Directional derivatives and gradient vectors (Sect. 14.5).

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- ▶ Properties of the the gradient vector.

## Directional derivative and partial derivatives.

**Remark:** The directional derivative  $(D_{\mathbf{u}}f)_{P_0}$  is the derivative of  $f$  along the line  $\mathbf{r}(t) = \langle x_0, y_0 \rangle + \mathbf{u} t$ .



### Theorem

If the function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is differentiable at  $P_0 = (x_0, y_0)$  and  $\mathbf{u} = \langle u_x, u_y \rangle$  is a unit vector, then

$$(D_{\mathbf{u}}f)_{P_0} = f_x(x_0, y_0) u_x + f_y(x_0, y_0) u_y.$$

## Directional derivative and partial derivatives.

### Proof.

The line  $\mathbf{r}(t) = \langle x_0, y_0 \rangle + \langle u_x, u_y \rangle t$  has parametric equations:  
 $x(t) = x_0 + u_x t$  and  $y(t) = y_0 + u_y t$ ;

Denote  $f$  evaluated along the line as  $\hat{f}(t) = f(x(t), y(t))$ .

Now, on the one hand,  $\hat{f}'(0) = (D_{\mathbf{u}}f)_{P_0}$ , since

$$\begin{aligned}\hat{f}'(0) &= \lim_{t \rightarrow 0} \frac{1}{t} [\hat{f}(t) - \hat{f}(0)], \\ &= \lim_{t \rightarrow 0} \frac{1}{t} [f(x_0 + u_x t, y_0 + u_y t) - f(x_0, y_0)], \\ &= D_{\mathbf{u}}f(x_0, y_0).\end{aligned}$$

On the other hand, the chain rule implies:

$$\hat{f}'(0) = f_x(x_0, y_0) x'(0) + f_y(x_0, y_0) y'(0).$$

Therefore,  $(D_{\mathbf{u}}f)_{P_0} = f_x(x_0, y_0) u_x + f_y(x_0, y_0) u_y$ . □

## Directional derivative and partial derivatives.

### Example

Compute the directional derivative of  $f(x, y) = \sin(x + 3y)$  at the point  $P_0 = (4, 3)$  in the direction of vector  $\mathbf{v} = \langle 1, 2 \rangle$ .

**Solution:** We need to find a unit vector in the direction of  $\mathbf{v}$ .

Such vector is  $\mathbf{u} = \frac{\mathbf{v}}{|\mathbf{v}|} \Rightarrow \mathbf{u} = \frac{1}{\sqrt{5}} \langle 1, 2 \rangle$ .

We now use the formula  $(D_{\mathbf{u}}f)_{P_0} = f_x(x_0, y_0) u_x + f_y(x_0, y_0) u_y$ .

That is,  $(D_{\mathbf{u}}f)_{P_0} = \cos(x_0 + 3y_0)(1/\sqrt{5}) + 3 \cos(x_0 + 3y_0)(2/\sqrt{5})$ .

Equivalently,  $(D_{\mathbf{u}}f)_{P_0} = (7/\sqrt{5}) \cos(x_0 + 3y_0)$ .

Then,  $(D_{\mathbf{u}}f)_{P_0} = (7/\sqrt{5}) \cos(10)$ . ◁

## Directional derivatives and gradient vectors (Sect. 14.5).

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- ▶ **Directional derivative of functions of three variables.**
- ▶ The gradient vector and directional derivatives.
- ▶ Properties of the the gradient vector.

## Directional derivative of functions of three variables.

### Definition

The *directional derivative* of the function  $f : D \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  at the point  $P_0 = (x_0, y_0, z_0) \in D$  in the direction of a unit vector  $\mathbf{u} = \langle u_x, u_y, u_z \rangle$  is given by

$$(D_{\mathbf{u}}f)_{P_0} = \lim_{t \rightarrow 0} \frac{1}{t} [f(x_0 + u_x t, y_0 + u_y t, z_0 + u_z t) - f(x_0, y_0, z_0)],$$

if the limit exists.

### Theorem

If the function  $f : D \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  is differentiable at  $P_0 = (x_0, y_0, z_0)$  and  $\mathbf{u} = \langle u_x, u_y, u_z \rangle$  is a unit vector, then

$$(D_{\mathbf{u}}f)_{P_0} = f_x(x_0, y_0, z_0) u_x + f_y(x_0, y_0, z_0) u_y + f_z(x_0, y_0, z_0) u_z.$$

## Directional derivative of functions of three variables.

### Example

Find  $(D_{\mathbf{u}}f)_{P_0}$  for  $f(x, y, z) = x^2 + 2y^2 + 3z^2$  at the point  $P_0 = (3, 2, 1)$  along the direction given by  $\mathbf{v} = \langle 2, 1, 1 \rangle$ .

**Solution:** We first find a unit vector along  $\mathbf{v}$ ,

$$\mathbf{u} = \frac{\mathbf{v}}{|\mathbf{v}|} \Rightarrow \mathbf{u} = \frac{1}{\sqrt{6}} \langle 2, 1, 1 \rangle.$$

Then,  $(D_{\mathbf{u}}f)$  is given by  $(D_{\mathbf{u}}f) = (2x)u_x + (4y)u_y + (6z)u_z$ .

We conclude,  $(D_{\mathbf{u}}f)_{P_0} = (6)\frac{2}{\sqrt{6}} + (8)\frac{1}{\sqrt{6}} + (6)\frac{1}{\sqrt{6}}$ ,

that is,  $(D_{\mathbf{u}}f)_{P_0} = \frac{26}{\sqrt{6}}$ .

◁

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## The gradient vector and directional derivatives.

**Remark:** The directional derivative of a function can be written in terms of a dot product.

- ▶ In the case of 2 variable functions:  $D_{\mathbf{u}}f = f_x u_x + f_y u_y$

$$D_{\mathbf{u}}f = (\nabla f) \cdot \mathbf{u}, \quad \text{with} \quad \nabla f = \langle f_x, f_y \rangle.$$

- ▶ In the case of 3 variable functions:  $D_{\mathbf{u}}f = f_x u_x + f_y u_y + f_z u_z$ ,

$$D_{\mathbf{u}}f = (\nabla f) \cdot \mathbf{u}, \quad \text{with} \quad \nabla f = \langle f_x, f_y, f_z \rangle.$$

## The gradient vector and directional derivatives.

### Definition

The *gradient vector* of a differentiable function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  at any point  $(x, y) \in D$  is the vector  $\nabla f = \langle f_x, f_y \rangle$ .

The *gradient vector* of a differentiable function  $f : D \subset \mathbb{R}^3 \rightarrow \mathbb{R}$  at any point  $(x, y, z) \in D$  is the vector  $\nabla f = \langle f_x, f_y, f_z \rangle$ .

### Notation:

- ▶ For two variable functions:  $\nabla f = f_x \mathbf{i} + f_y \mathbf{j}$ .
- ▶ For three variable functions:  $\nabla f = f_x \mathbf{i} + f_y \mathbf{j} + f_z \mathbf{k}$ .

### Theorem

If  $f : D \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , with  $n = 2, 3$ , is a differentiable function and  $\mathbf{u}$  is a unit vector, then,

$$D_{\mathbf{u}}f = (\nabla f) \cdot \mathbf{u}.$$

## The gradient vector and directional derivatives.

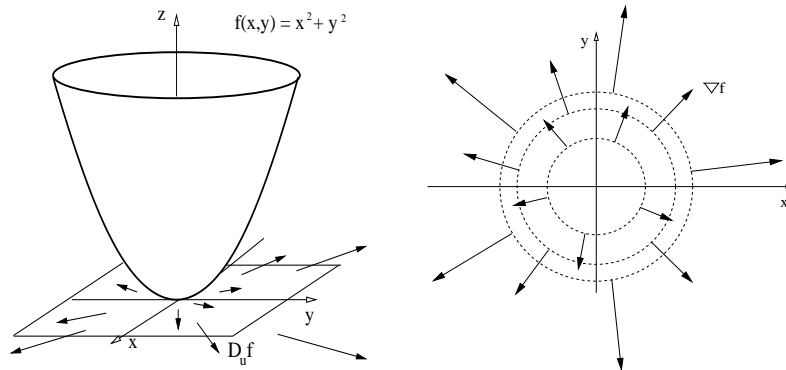
### Example

Find the gradient vector at any point in the domain of the function  $f(x, y) = x^2 + y^2$ .

**Solution:** The gradient is  $\nabla f = \langle f_x, f_y \rangle$ , that is,  $\nabla f = \langle 2x, 2y \rangle$ .  $\triangleleft$

### Remark:

$\nabla f = 2\mathbf{r}$ ,  
with  
 $\mathbf{r} = \langle x, y \rangle$ .



## Directional derivatives and gradient vectors (Sect. 14.5).

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## Properties of the the gradient vector.

**Remark:** If  $\theta$  is the angle between  $\nabla f$  and  $\mathbf{u}$ , then holds

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

The formula above implies:

- ▶ The function  $f$  increases the most rapidly when  $\mathbf{u}$  is in the direction of  $\nabla f$ , that is,  $\theta = 0$ . The maximum increase rate of  $f$  is  $|\nabla f|$ .
- ▶ The function  $f$  decreases the most rapidly when  $\mathbf{u}$  is in the direction of  $-\nabla f$ , that is,  $\theta = \pi$ . The maximum decrease rate of  $f$  is  $-|\nabla f|$ .
- ▶ The function  $f$  does not change along level curve or surfaces, that is,  $D_{\mathbf{u}}f = 0$ . Therefore,  $\nabla f$  is perpendicular to the level curves or level surfaces.

## Properties of the the gradient vector.

### Example

Find the direction of maximum increase of the function  $f(x, y) = x^2/4 + y^2/9$  at an arbitrary point  $(x, y)$ , and also at the points  $(1, 0)$  and  $(0, 1)$ .

**Solution:** The direction of maximum increase of  $f$  is the direction of its gradient vector:

$$\nabla f = \left\langle \frac{x}{2}, \frac{2y}{9} \right\rangle.$$

At the points  $(1, 0)$  and  $(0, 1)$  we obtain, respectively,

$$\nabla f = \left\langle \frac{1}{2}, 0 \right\rangle. \quad \nabla f = \left\langle 0, \frac{2}{9} \right\rangle.$$

## Properties of the the gradient vector.

### Example

Given the function  $f(x, y) = x^2/4 + y^2/9$ , find the equation of a line tangent to a level curve  $f(x, y) = 1$  at the point  $P_0 = (1, -3\sqrt{3}/2)$ .

**Solution:** We first verify that  $P_0$  belongs to the level curve  $f(x, y) = 1$ . This is the case, since

$$\frac{1}{4} + \frac{(9)(3)}{4} \frac{1}{9} = 1.$$

The equation of the line we look for is

$$\mathbf{r}(t) = \left\langle 1, -\frac{3\sqrt{3}}{2} \right\rangle + t \langle v_x, v_y \rangle,$$

where  $\mathbf{v} = \langle v_x, v_y \rangle$  is tangent to the level curve  $f(x, y) = 1$  at  $P_0$ .

## Properties of the the gradient vector.

### Example

Given the function  $f(x, y) = x^2/4 + y^2/9$ , find the equation of a line tangent to a level curve  $f(x, y) = 1$  at the point  $P_0 = (1, -3\sqrt{3}/2)$ .

**Solution:** Therefore,  $\mathbf{v} \perp \nabla f$  at  $P_0$ . Since,

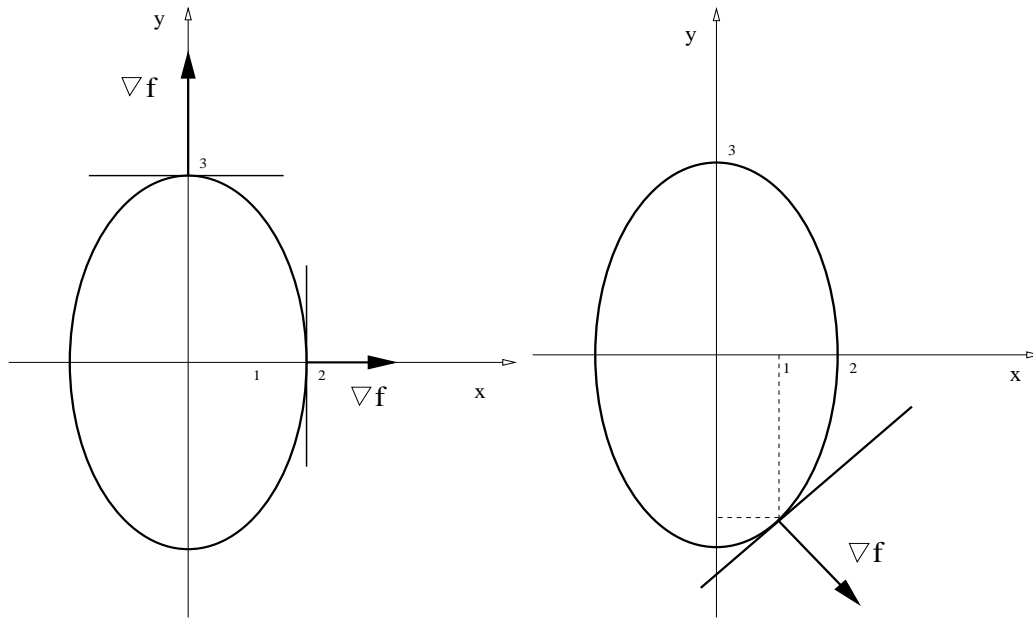
$$\nabla f = \left\langle \frac{x}{2}, \frac{2y}{9} \right\rangle \Rightarrow (\nabla f)_{P_0} = \left\langle \frac{1}{2}, -\frac{2 \cdot 3\sqrt{3}}{9} \right\rangle = \left\langle \frac{1}{2}, -\frac{1}{\sqrt{3}} \right\rangle.$$

Therefore,

$$0 = \mathbf{v} \cdot (\nabla f)_{P_0} \Rightarrow \frac{1}{2} v_x = \frac{1}{\sqrt{3}} v_y \Rightarrow \mathbf{v} = \langle 2, \sqrt{3} \rangle.$$

The line is  $\mathbf{r}(t) = \left\langle 1, -\frac{3\sqrt{3}}{2} \right\rangle + t \langle 2, \sqrt{3} \rangle.$  ◁

## Properties of the the gradient vector.



$$\nabla f = \left\langle \frac{1}{2}, 0 \right\rangle, \quad \nabla f = \left\langle 0, \frac{2}{9} \right\rangle, \quad \mathbf{r}(t) = \left\langle 1, -\frac{3\sqrt{3}}{2} \right\rangle + t \langle 2, \sqrt{3} \rangle.$$

## Further properties of the the gradient vector.

### Theorem

If  $f, g$  are differentiable scalar valued vector functions,  $g \neq 0$ , and  $k \in \mathbb{R}$  any constant, then holds,

1.  $\nabla(kf) = k(\nabla f)$ ;
2.  $\nabla(f \pm g) = \nabla f \pm \nabla g$ ;
3.  $\nabla(fg) = (\nabla f)g + f(\nabla g)$ ;
4.  $\nabla\left(\frac{f}{g}\right) = \frac{(\nabla f)g - f(\nabla g)}{g^2}$ .

## Tangent planes and linear approximations (Sect. 14.6).

- ▶ Review: Differentiable functions of two variables.
- ▶ The tangent plane to the graph of a function.
- ▶ The linear approximation of a differentiable function.
- ▶ Bounds for the error of a linear approximation.
- ▶ The differential of a function.
  - ▶ Review: Scalar functions of one variable.
  - ▶ Scalar functions of more than one variable.

## Review: Differentiable functions of two variables.

### Definition

Given a function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  and an interior point  $(x_0, y_0) \in D$ , let  $L$  be the linear function

$$L(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0).$$

The function  $f$  is called *differentiable at  $(x_0, y_0)$*  iff the function  $f$  is approximated by the linear function  $L$  near  $(x_0, y_0)$ , that is,

$$f(x, y) = L(x, y) + \epsilon_1(x - x_0) + \epsilon_2(y - y_0)$$

where the functions  $\epsilon_1$  and  $\epsilon_2 \rightarrow 0$  as  $(x, y) \rightarrow (x_0, y_0)$ .

### Theorem

*If the partial derivatives  $f_x$  and  $f_y$  of a function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  are continuous in an open region  $R \subset D$ , then  $f$  is differentiable in  $R$ .*

## Review: Differentiable functions of two variables.

### Example

Show that the function  $f(x, y) = x^2 + y^2$  is differentiable for all  $(x, y) \in \mathbb{R}^2$ . Furthermore, find the linear function  $L$ , mentioned in the definition of a differentiable function, at the point  $(1, 2)$ .

**Solution:** The partial derivatives of  $f$  are given by  $f_x(x, y) = 2x$  and  $f_y(x, y) = 2y$ , which are continuous functions. Therefore, the function  $f$  is differentiable. The linear function  $L$  at  $(1, 2)$  is

$$L(x, y) = f_x(1, 2)(x - 1) + f_y(1, 2)(y - 2) + f(1, 2).$$

That is, we need three numbers to find the linear function  $L$ :  $f_x(1, 2)$ ,  $f_y(1, 2)$ , and  $f(1, 2)$ . These numbers are:

$$f_x(1, 2) = 2, \quad f_y(1, 2) = 4, \quad f(1, 2) = 5.$$

Therefore,  $L(x, y) = 2(x - 1) + 4(y - 2) + 5$ . ◁

## Tangent planes and linear approximations (Sect. 14.6).

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## The tangent plane to the graph of a function.

Remark:

The function  $L(x, y) = 2(x - 1) + 4(y - 2) + 5$  is a plane in  $\mathbb{R}^3$ . We usually write down the equation of a plane using the notation  $z = L(x, y)$ , that is,  $z = 2(x - 1) + 4(y - 2) + 5$ , or equivalently

$$2(x - 1) + 4(y - 2) - (z - 5) = 0.$$

This is a plane passing through  $\tilde{P}_0 = (1, 2, 5)$  with normal vector  $\mathbf{n} = \langle 2, 4, -1 \rangle$ . Analogously, the function

$$L(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0)$$

is a plane in  $\mathbb{R}^3$ . Using the notation  $z = L(x, y)$  we obtain

$$f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) - (z - f(x_0, y_0)) = 0.$$

This is a plane passing through  $\tilde{P}_0 = (x_0, y_0, f(x_0, y_0))$  with normal vector  $\mathbf{n} = \langle f_x(x_0, y_0), f_y(x_0, y_0), -1 \rangle$ .

## The tangent plane to the graph of a function.

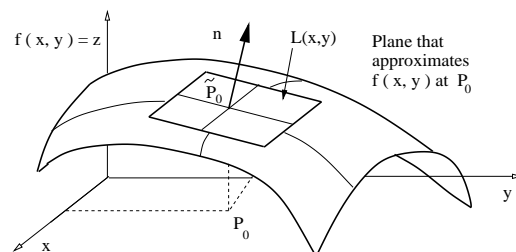
Theorem

The plane tangent to the graph of a differentiable function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  at the point  $(x_0, y_0)$  is given by

$$L(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0).$$

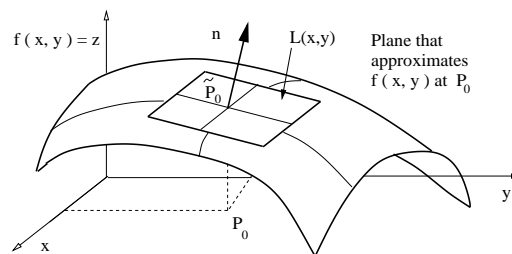
Proof

The plane contains the point  $\tilde{P}_0 = (x_0, y_0, f(x_0, y_0))$ . We only need to find its normal vector  $\mathbf{n}$ .



## The tangent plane to the graph of a function.

The vector  $\mathbf{n}$  normal to the plane  $L(x, y)$  is a vector perpendicular to the surface  $z = f(x, y)$  at  $P_0 = (x_0, y_0)$ .



This surface is the level surface  $F(x, y, z) = 0$  of the function  $F(x, y, z) = f(x, y) - z$ . A vector normal to this level surface is its gradient  $\nabla F$ . That is,  $\nabla F = \langle F_x, F_y, F_z \rangle = \langle f_x, f_y, -1 \rangle$ .

Therefore, the normal to the tangent plane  $L(x, y)$  at the point  $P_0$  is  $\mathbf{n} = \langle f_x(x_0, y_0), f_y(x_0, y_0), -1 \rangle$ . Recall that the plane contains the point  $\tilde{P}_0 = (x_0, y_0, f(x_0, y_0))$ . The equation for the plane is

$$f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) - (z - f(x_0, y_0)) = 0.$$

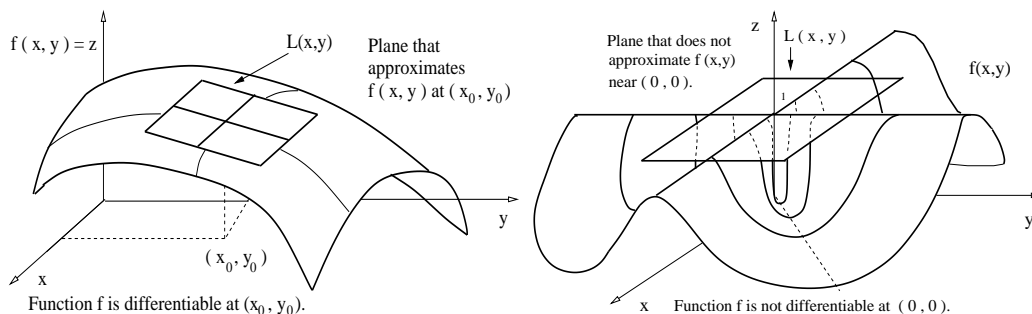
□

## The tangent plane to the graph of a function.

**Summary:** We have shown that the linear  $L$  given by

$$L(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0)$$

is the plane tangent to the graph of  $f$  at  $(x_0, y_0)$ .



**Remark:** The graph of a differentiable function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is approximated by the tangent plane  $L$  at every point in  $D$ .

## The tangent plane to the graph of a function.

### Example

Show that  $f(x, y) = \arctan(x + 2y)$  is differentiable and find the plane tangent to  $f(x, y)$  at  $(1, 0)$ .

**Solution:** The partial derivatives of  $f$  are given by

$$f_x(x, y) = \frac{1}{1 + (x + 2y)^2}, \quad f_y(x, y) = \frac{2}{1 + (x + 2y)^2}.$$

These functions are continuous in  $\mathbb{R}^2$ , so  $f(x, y)$  is differentiable at every point in  $\mathbb{R}^2$ . The plane  $L(x, y)$  at  $(1, 0)$  is given by

$$L(x, y) = f_x(1, 0)(x - 1) + f_y(1, 0)(y - 0) + f(1, 0),$$

where  $f(1, 0) = \arctan(1) = \pi/4$ ,  $f_x(1, 0) = 1/2$ ,  $f_y(1, 0) = 1$ .

Then,  $L(x, y) = \frac{1}{2}(x - 1) + y + \frac{\pi}{4}$ . ◁

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## The linear approximation of a differentiable function.

### Definition

The *linear approximation* of a differentiable function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  at the point  $(x_0, y_0) \in D$  is the plane

$$L(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0).$$

### Example

Find the linear approximation of  $f = \sqrt{17 - x^2 - 4y^2}$  at  $(2, 1)$ .

**Solution:**  $L(x, y) = f_x(2, 1)(x - 2) + f_y(2, 1)(y - 1) + f(2, 1)$ .

We need three numbers:  $f(2, 1)$ ,  $f_x(2, 1)$ , and  $f_y(2, 1)$ .

These are:  $f(2, 1) = 3$ ,  $f_x(2, 1) = -2/3$ , and  $f_y(2, 1) = -4/3$ .

Then the plane is given by  $L(x, y) = -\frac{2}{3}(x - 2) - \frac{4}{3}(y - 1) + 3$ . ◁

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- ▶ Review: Differentiable functions of two variables.
- ▶ The tangent plane to the graph of a function.
- ▶ The linear approximation of a differentiable function.
- ▶ **Bounds for the error of a linear approximation.**
- ▶ The differential of a function.
  - ▶ Review: Scalar functions of one variable.
  - ▶ Scalar functions of more than one variable.

## Bounds for the error of a linear approximation.

### Theorem

Assume that the function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  has first and second partial derivatives continuous on an open set containing a rectangular region  $R \subset D$  centered at the point  $(x_0, y_0)$ . If  $M \in \mathbb{R}$  is the upper bound for  $|f_{xx}|$ ,  $|f_{yy}|$ , and  $|f_{xy}|$  in  $R$ , then the error  $E(x, y) = f(x, y) - L(x, y)$  satisfies the inequality

$$|E(x, y)| \leq \frac{1}{2} M (|x - x_0| + |y - y_0|)^2,$$

where  $L(x, y)$  is the linearization of  $f$  at  $(x_0, y_0)$ , that is,

$$L(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0).$$

## Bounds for the error of a linear approximation.

### Example

Find an upper bound for the error in the linear approximation of  $f(x, y) = x^2 + y^2$  at the point  $(1, 2)$  over the rectangle

$$R = \{(x, y) \in \mathbb{R}^2 : |x - 1| < 0.1, \quad |y - 2| < 0.1\}$$

**Solution:** The second derivatives of  $f$  are  $f_{xx} = 2$ ,  $f_{yy} = 2$ ,  $f_{xy} = 0$ .

Therefore, we can take  $M = 2$ .

Then the formula  $|E(x, y)| \leq \frac{1}{2} M (|x - x_0| + |y - y_0|)^2$ , implies

$$|E(x, y)| \leq (|x - 1| + |y - 2|)^2 < (0.1 + 0.1)^2 = 0.04,$$

that is  $|E(x, y)| < 0.04$ . Since  $f(1, 2) = 5$ , the percentage relative error  $100 E(x, y)/f(1, 2)$  is bounded by **0.8%**  $\triangleleft$

## Tangent planes and linear approximations (Sect. 14.6).

- ▶ Review: Differentiable functions of two variables.
- ▶ The tangent plane to the graph of a function.
- ▶ The linear approximation of a differentiable function.
- ▶ Bounds for the error of a linear approximation.
- ▶ **The differential of a function.**
  - ▶ Review: Scalar functions of one variable.
  - ▶ Scalar functions of more than one variable.

## Review: Differential of functions of one variable.

### Definition

The *differential at  $x_0 \in D$*  of a differentiable function  $f : D \subset \mathbb{R} \rightarrow \mathbb{R}$  is the linear function

$$df(x) = L(x) - f(x_0).$$

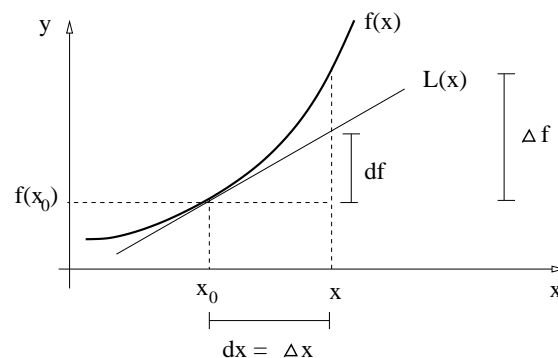
**Remark:** The linear approximation of  $f(x)$  at  $x_0$  is the line given by  $L(x) = f'(x_0)(x - x_0) + f(x_0)$ .

Therefore

$$df(x) = f'(x_0)(x - x_0).$$

Denoting  $dx = x - x_0$ ,

$$df = f'(x_0) dx.$$



## Differential of functions of more than one variable.

### Definition

The *differential at*  $(x_0, y_0) \in D$  of a differentiable function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is the linear function

$$df(x, y) = L(x, y) - f(x_0, y_0).$$

**Remark:** The linear approximation of  $f(x, y)$  at  $(x_0, y_0)$  is the plane  $L(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0) + f(x_0, y_0)$ .

Therefore  $df(x, y) = f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0)$ .

Denoting  $dx = x - x_0$  and  $dy = (y - y_0)$  we obtain the usual expression

$$df = f_x(x_0, y_0) dx + f_y(x_0, y_0) dy.$$

Therefore,  $df$  and  $L$  are similar concepts: The linear approximation of a differentiable function  $f$ .

## Differential of functions of more than one variable.

### Example

Compute the  $df$  of the function  $f(x, y) = \ln(1 + x^2 + y^2)$  at the point  $(1, 1)$ . Evaluate this  $df$  for  $dx = 0.1$ ,  $dy = 0.2$ .

**Solution:** The differential of  $f$  at  $(x_0, y_0)$  is given by

$$df = f_x(x_0, y_0)dx + f_y(x_0, y_0)dy.$$

The partial derivatives  $f_x$  and  $f_y$  are given by

$$f_x(x, y) = \frac{2x}{1 + x^2 + y^2}, \quad f_y(x, y) = \frac{2y}{1 + x^2 + y^2}.$$

Therefore,  $f_x(1, 1) = 2/3 = f_y(1, 1)$ . Then  $df = \frac{2}{3} dx + \frac{2}{3} dy$ . Evaluating this differential at  $dx = 0.1$  and  $dy = 0.2$  we obtain

$$df = \frac{2}{3} \frac{1}{10} + \frac{2}{3} \frac{2}{10} = \frac{2}{3} \frac{3}{10} \Rightarrow df = \frac{1}{5}.$$

## Differential of functions of more than one variable.

### Example

Use differentials to estimate the amount of aluminum needed to build a closed cylindrical can with internal diameter of  $8\text{cm}$  and height of  $12\text{cm}$  if the aluminum is  $0.04\text{cm}$  thick.

### Solution:

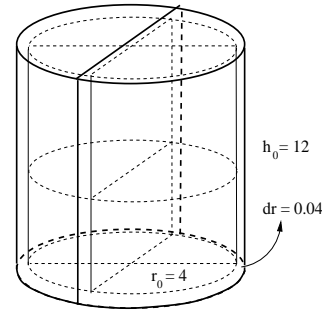
The data of the problem is:  $h_0 = 12\text{cm}$ ,  
 $r_0 = 4\text{cm}$ ,  $dr = 0.04\text{cm}$  and  $dh = 0.08\text{cm}$ .

The function to consider is the mass of the cylinder,  $M = \rho V$ , where  $\rho = 2.7\text{gr}/\text{cm}^3$  is the aluminum density and  $V$  is the volume of the cylinder,

$$V(r, h) = \pi r^2 h.$$

The metal to build the can is given by

$$\Delta M = \rho [V(r + dr, h + dh) - V(r, h)], \quad (\text{recall } dh = 2dr.)$$



## Differential of functions of more than one variable.

### Example

Use differentials to estimate the amount of aluminum needed to build a closed cylindrical can with internal diameter of  $8\text{cm}$  and height of  $12\text{cm}$  if the aluminum is  $0.04\text{cm}$  thick.

**Solution:** The metal to build the can is given by

$$\Delta M = \rho [V(r + dr, h + dh) - V(r, h)],$$

A linear approximation to  $\Delta V = V(r + dr, h + dh) - V(r, h)$  is  $dV = V_r dr + V_h dh$ , that is,

$$dV = V_r(r_0, h_0)dr + V_h(r_0, h_0)dh.$$

Since  $V(r, h) = \pi r^2 h$ , we obtain  $dV = 2\pi r_0 h_0 dr + \pi r_0^2 dh$ .

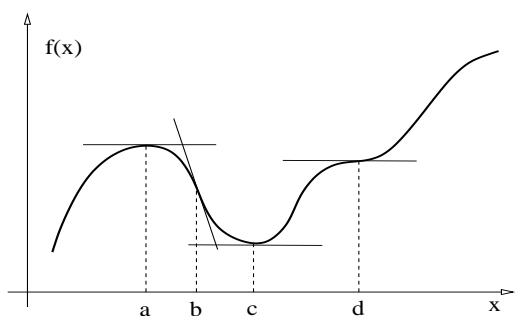
Therefore,  $dV = 16.1\text{ cm}^3$ . Since  $dM = \rho dV$ , a linear estimate for the aluminum needed to build the can is  $dM = 43.47\text{ gr}$ .  $\triangleleft$

## Local and absolute extrema, saddle points (Sect. 14.7).

- ▶ Review: Local extrema for functions of one variable.
- ▶ Definition of local extrema.
- ▶ Characterization of local extrema.
  - ▶ First derivative test.
  - ▶ Second derivative test.
- ▶ Absolute extrema of a function in a domain.

## Review: Local extrema for functions of one variable.

Recall: Main results on local extrema for  $f(x)$ :



at	$f$	$f'$	$f''$
$a$	max.	$0$	$< 0$
$b$	infl.	$\neq 0$	$\pm 0 \mp$
$c$	min.	$0$	$> 0$
$d$	infl.	$= 0$	$\pm 0 \mp$

**Remarks:** Assume that  $f$  is twice continuously differentiable.

- ▶ If  $x_0$  is local maximum or minimum of  $f$ , then  $f'(x_0) = 0$ .
- ▶ If  $f'(x_0) = 0$  then  $x_0$  is a critical point of  $f$ , that is,  $x_0$  is a maximum or a minimum or an inflection point.
- ▶ The second derivative test determines whether a critical point is a maximum, minimum or an inflection point.

## Local and absolute extrema, saddle points (Sect. 14.7).

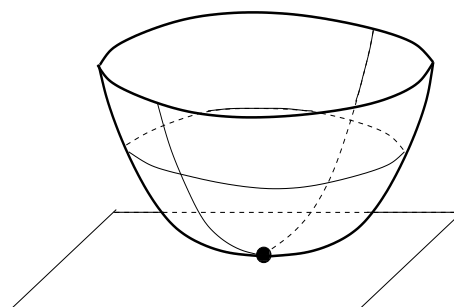
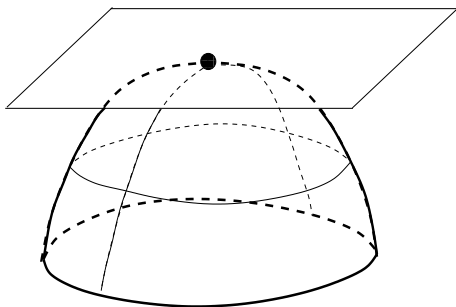
- ▶ Review: Local extrema for functions of one variable.
- ▶ **Definition of local extrema.**
- ▶ Characterization of local extrema.
  - ▶ First derivative test.
  - ▶ Second derivative test.
- ▶ Absolute extrema of a function in a domain.

## Definition of local extrema for functions of two variables.

### Definition

A function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  has a *local maximum* at the point  $(a, b) \in D$  iff holds that  $f(x, y) \leq f(a, b)$  for every point  $(x, y)$  in a neighborhood of  $(a, b)$ .

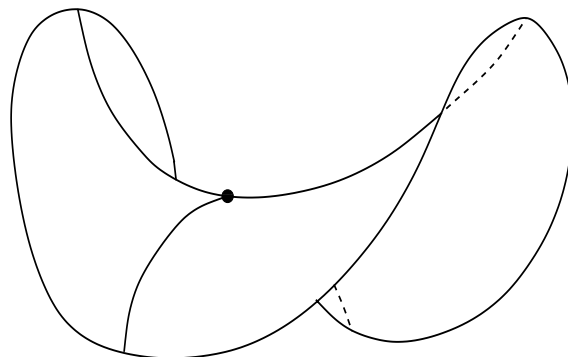
A function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  has a *local minimum* at the point  $(a, b) \in D$  iff holds that  $f(x, y) \geq f(a, b)$  for every point  $(x, y)$  in a neighborhood of  $(a, b)$ .



## Definition of local extrema for functions of two variables.

### Definition

A differentiable function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  has a *saddle point* at an interior point  $(a, b) \in D$  iff in every open disk in  $D$  centered at  $(a, b)$  there always exist points  $(x, y)$  where  $f(x, y) > f(a, b)$  and other points  $(x, y)$  where  $f(x, y) < f(a, b)$ .



## Local and absolute extrema, saddle points (Sect. 14.7).

- ▶ Review: Local extrema for functions of one variable.
- ▶ Definition of local extrema.
- ▶ **Characterization of local extrema.**
  - ▶ First derivative test.
  - ▶ Second derivative test.
- ▶ Absolute extrema of a function in a domain.



## Characterization of local extrema.

First derivative test.

### Theorem

If a differentiable function  $f$  has a local maximum or minimum at  $(a, b)$  then holds  $(\nabla f)|_{(a,b)} = \langle 0, 0 \rangle$ .

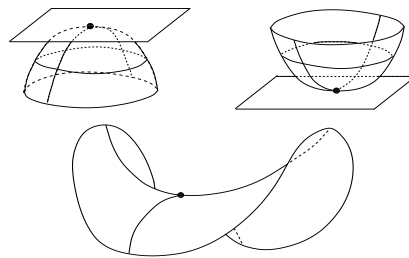
**Remark:** The tangent plane at a local extremum is horizontal, since its normal vector is  $\mathbf{n} = \langle f_x, f_y, -1 \rangle = \langle 0, 0, -1 \rangle$ .

### Definition

The interior point  $(a, b) \in D$  of a differentiable function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is a *critical point* of  $f$  iff  $(\nabla f)|_{(a,b)} = \langle 0, 0 \rangle$ .

### Remark:

Critical points include local maxima, local minima, and saddle points.



## Characterization of local extrema.

First derivative test.

### Example

Find the critical points of the function  $f(x, y) = -x^2 - y^2$

**Solution:** The critical points are the points where  $\nabla f$  vanishes. Since  $\nabla f = \langle -2x, -2y \rangle$ , the only solution to  $\nabla f = \langle 0, 0 \rangle$  is  $x = 0$ ,  $y = 0$ . That is,  $(a, b) = (0, 0)$ .  $\triangleleft$

**Remark:** Since  $f(x, y) \leq 0$  for all  $(x, y) \in \mathbb{R}^2$  and  $f(0, 0) = 0$ , then the point  $(0, 0)$  must be a local maximum of  $f$ .

### Example

Find the critical points of the function  $f(x, y) = x^2 - y^2$

**Solution:** Since  $\nabla f = \langle 2x, -2y \rangle$ , the only solution to  $\nabla f = \langle 0, 0 \rangle$  is  $x = 0$ ,  $y = 0$ . That is, we again obtain  $(a, b) = (0, 0)$ .  $\triangleleft$

## Characterization of local extrema.

Second derivative test.

### Theorem

Let  $(a, b)$  be a critical point of  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$ , that is,  $(\nabla f)|_{(a,b)} = \langle 0, 0 \rangle$ . Assume that  $f$  has continuous second derivatives in an open disk in  $D$  with center in  $(a, b)$  and denote

$$D = f_{xx}(a, b) f_{yy}(a, b) - [f_{xy}(a, b)]^2.$$

Then, the following statements hold:

- ▶ If  $D > 0$  and  $f_{xx}(a, b) > 0$ , then  $f(a, b)$  is a local minimum.
- ▶ If  $D > 0$  and  $f_{xx}(a, b) < 0$ , then  $f(a, b)$  is a local maximum.
- ▶ If  $D < 0$ , then  $f(a, b)$  is a saddle point.
- ▶ If  $D = 0$  the test is inconclusive.

**Notation:** The number  $D$  is called the **discriminant** of  $f$  at  $(a, b)$ .

## Characterization of local extrema.

Second derivative test.

### Example

Find the local extrema of  $f(x, y) = y^2 - x^2$  and determine whether they are local maximum, minimum, or saddle points.

**Solution:** We first find the critical points:

$$\nabla f = \langle -2x, 2y \rangle \Rightarrow (\nabla f)|_{(a,b)} = \langle 0, 0 \rangle \text{ iff } (a, b) = (0, 0).$$

The only critical point is  $(a, b) = (0, 0)$ .

We need to compute  $D = f_{xx}(a, b) f_{yy}(a, b) - [f_{xy}(a, b)]^2$ .

Since  $f_{xx}(0, 0) = -2$ ,  $f_{yy}(0, 0) = 2$ , and  $f_{xy}(0, 0) = 0$ , we get

$$D = (-2)(2) = -4 < 0 \Rightarrow \text{saddle point at } (0, 0).$$



## Characterization of local extrema.

Second derivative test.

### Example

Is the point  $(a, b) = (0, 0)$  a local extrema of  $f(x, y) = y^2x^2$ ?

**Solution:** We first verify that  $(0, 0)$  is a critical point of  $f$ :

$$\nabla f(x, y) = \langle 2xy^2, 2yx^2 \rangle, \quad \Rightarrow \quad (\nabla f)|_{(0,0)} = \langle 0, 0 \rangle,$$

therefore,  $(0, 0)$  is a critical point.

**Remark:** The whole axes  $x = 0$  and  $y = 0$  are critical points of  $f$ .

We need to compute  $D = f_{xx}(a, b) f_{yy}(a, b) - [f_{xy}(a, b)]^2$ .

Since  $f_{xx}(x, y) = 2y^2$ ,  $f_{yy}(x, y) = 2x^2$ , and  $f_{xy}(x, y) = 4xy$ ,

we obtain  $f_{xx}(0, 0) = 0$ ,  $f_{yy}(0, 0) = 0$ , and  $f_{xy}(0, 0) = 0$ ,

hence  $D = 0$  and **the test is inconclusive.**  $\triangleleft$

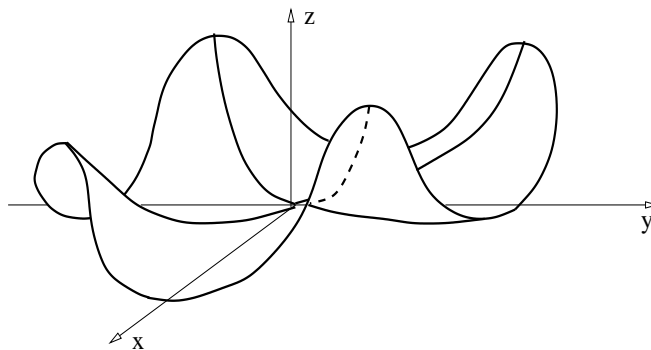
## Characterization of local extrema.

Second derivative test.

### Example

Is the point  $(a, b) = (0, 0)$  a local extrema of  $f(x, y) = y^2x^2$ ?

**Solution:** From the graph of  $f = x^2y^2$  is simple to see that  $(0, 0)$  is a **local minimum:** (also a global minimum.)  $\triangleleft$



## Local and absolute extrema, saddle points (Sect. 14.7).

- ▶ Review: Local extrema for functions of one variable.
- ▶ Definition of local extrema.
- ▶ Characterization of local extrema.
  - ▶ First derivative test.
  - ▶ Second derivative test.
- ▶ **Absolute extrema of a function in a domain.**

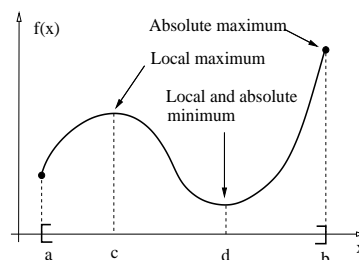
## Absolute extrema of a function in a domain.

### Definition

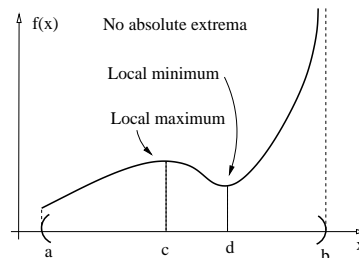
A function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  has an **absolute maximum** at the point  $(a, b) \in D$  iff  $f(x, y) \leq f(a, b)$  for all  $(x, y) \in D$ .

A function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  has an **absolute minimum** at the point  $(a, b) \in D$  iff  $f(x, y) \geq f(a, b)$  for all  $(x, y) \in D$ .

**Remark:** Local extrema need not be the absolute extrema.



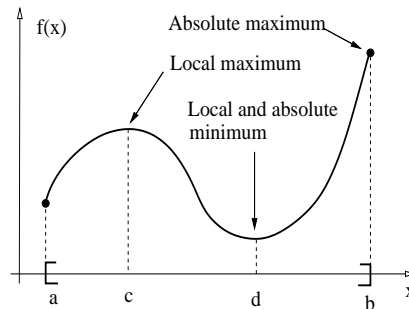
**Remark:** Absolute extrema may not be defined on open intervals.



## Review: Functions of one variable.

### Theorem

Every continuous functions  $f : [a, b] \subset \mathbb{R} \rightarrow \mathbb{R}$ , with  $a < b \in \mathbb{R}$  always has absolute extrema.



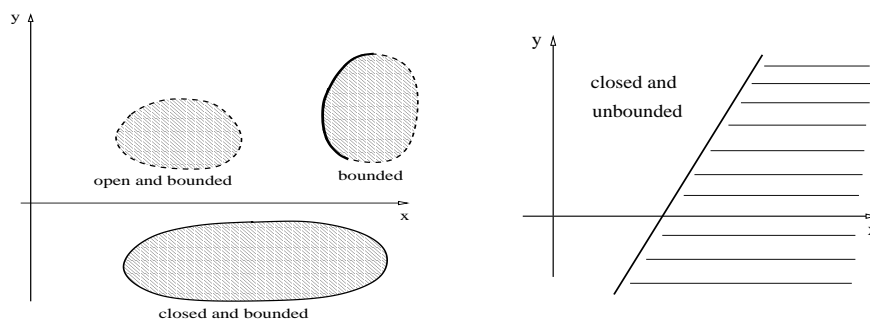
### Recall:

- ▶ Intervals  $[a, b]$  are bounded and closed sets in  $\mathbb{R}$ .
- ▶ The set  $[a, b]$  is closed, since the boundary points belong to the set, and it is bounded, since it does not extend to infinity.

## Recall: On open and closed sets in $\mathbb{R}^n$ .

### Definition

A set  $S \in \mathbb{R}^n$ , with  $n \in \mathbb{N}$ , is called *open* iff every point in  $S$  is an interior point. The set  $S$  is called *closed* iff  $S$  contains its boundary. A set  $S$  is called *bounded* iff  $S$  is contained in ball, otherwise  $S$  is called *unbounded*.



### Theorem

If  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  is continuous in a closed and bounded set  $D$ , then  $f$  has an absolute maximum and an absolute minimum in  $D$ .

## Absolute extrema on closed and bounded sets.

### Problem:

Find the absolute extrema of a function  $f : D \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  in a closed and bounded set  $D$ .

### Solution:

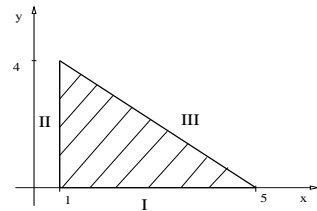
- (1) Find every critical point of  $f$  in the interior of  $D$  and evaluate  $f$  at these points.
- (2) Find the boundary points of  $D$  where  $f$  has local extrema, and evaluate  $f$  at these points.
- (3) Look at the list of values for  $f$  found in the previous two steps.

If  $f(x_0, y_0)$  is the biggest (smallest) value of  $f$  in the list above, then  $(x_0, y_0)$  is the absolute maximum (minimum) of  $f$  in  $D$ .

## Absolute extrema on closed and bounded sets.

### Example

Find the absolute extrema of the function  $f(x, y) = 3 + xy - x + 2y$  on the closed domain given in the Figure.



### Solution:

- (1) We find all critical points in the interior of the domain:

$$\nabla f = \langle (y - 1), (x + 2) \rangle = \langle 0, 0 \rangle \Rightarrow (x_0, y_0) = (-2, 1).$$

Since  $(-2, 1)$  does not belong to the domain, we discard it.

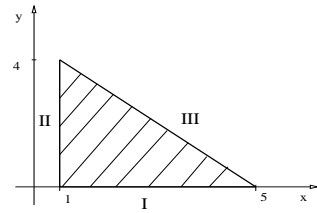
- (2) Three segments form the boundary of  $D$ :

**Boundary I:** The segment  $y = 0, x \in [1, 5]$ . We select the end points  $(1, 0)$ ,  $(5, 0)$ , and we record:  $f(1, 0) = 2$  and  $f(5, 0) = -2$ . We look for critical point on the interior of Boundary I: Since  $g(x) = f(x, 0) = 3 - x$ , so  $g' = -1 \neq 0$ . No critical points in the interior of Boundary I.

## Absolute extrema on closed and bounded sets.

### Example

Find the absolute extrema of the function  $f(x, y) = 3 + xy - x + 2y$  on the closed domain given in the Figure.



**Solution: Boundary II:** The segment  $x = 1, y \in [0, 4]$ . We select the end point  $(1, 4)$  and we record:  $f(1, 4) = 14$ .

We look for critical point on the interior of Boundary II: Since  $g(y) = f(1, y) = 3 + y - 1 + 2y = 2 + 3y$ , so  $g' = 3 \neq 0$ . **No critical points in the interior of Boundary II.**

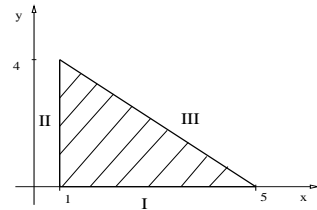
**Boundary III:** The segment  $y = -x + 5, x \in [1, 5]$ .

We look for critical point on the interior of Boundary III: Since  $g(x) = f(x, -x + 5) = 3 + x(-x + 5) - x + 2(-x + 5)$ . We obtain  $g(x) = -x^2 + 2x + 13$ , hence  $g'(x) = -2x + 2 = 0$  implies  $x = 1$ . So,  $y = 4$ , and we selected the point  $(1, 4)$ , which was already in our list. **No critical points in the interior of Boundary III.**

## Absolute extrema on closed and bounded sets.

### Example

Find the absolute extrema of the function  $f(x, y) = 3 + xy - x + 2y$  on the closed domain given in the Figure.



**Solution:**

(3) Our list of values is:

$$f(1, 0) = 2 \quad f(1, 4) = 14 \quad f(5, 0) = -2.$$

We conclude:

- ▶ Absolute maximum at  $(1, 4)$ ,
- ▶ Absolute minimum at  $(5, 0)$ .



## A maximization problem with a constraint.

### Example

Find the maximum volume of a closed rectangular box with a given surface area  $A_0$ .

**Solution:** This problem can be solved by finding the local maximum of an appropriate function  $f$ .

The function  $f$  is obtained as follows: Recall the functions volume and area of a rectangular box with vertex at  $(0, 0, 0)$  and sides  $x$ ,  $y$  and  $z$ :

$$V(x, y, z) = xyz, \quad A(x, y, z) = 2xy + 2xz + 2yz.$$

Since  $A(x, y, z) = A_0$ , we obtain  $z = \frac{A_0 - 2xy}{2(x + y)}$ , that is

$$f(x, y) = \frac{A_0xy - 2x^2y^2}{2(x + y)}.$$

## A maximization problem with a constraint.

### Example

Find the maximum volume of a closed rectangular box with a given surface area  $A_0$ .

**Solution:**

We must find the critical points of  $f(x, y) = \frac{A_0xy - 2x^2y^2}{2(x + y)}$ .

$$f_x = \frac{2A_0y^2 - 4x^2y^2 - 8xy^3}{4(x + y)^2}, \quad f_y = \frac{2A_0x^2 - 4x^2y^2 - 8yx^3}{4(x + y)^2}.$$

The conditions  $f_x = 0$  and  $f_y = 0$  and  $x \neq 0$ ,  $y \neq 0$  imply

$$\left. \begin{array}{l} A_0 = 2x^2 + 4xy, \\ A_0 = 2y^2 + 4xy, \end{array} \right\} \Rightarrow x = y. \quad \text{Recall } z = \frac{A_0 - 2xy}{2(x + y)},$$

so,  $z = \frac{A_0 - 2x^2}{4x} = y$ . Therefore,  $x_0 = y_0 = z_0 = \sqrt{A_0/6}$ .