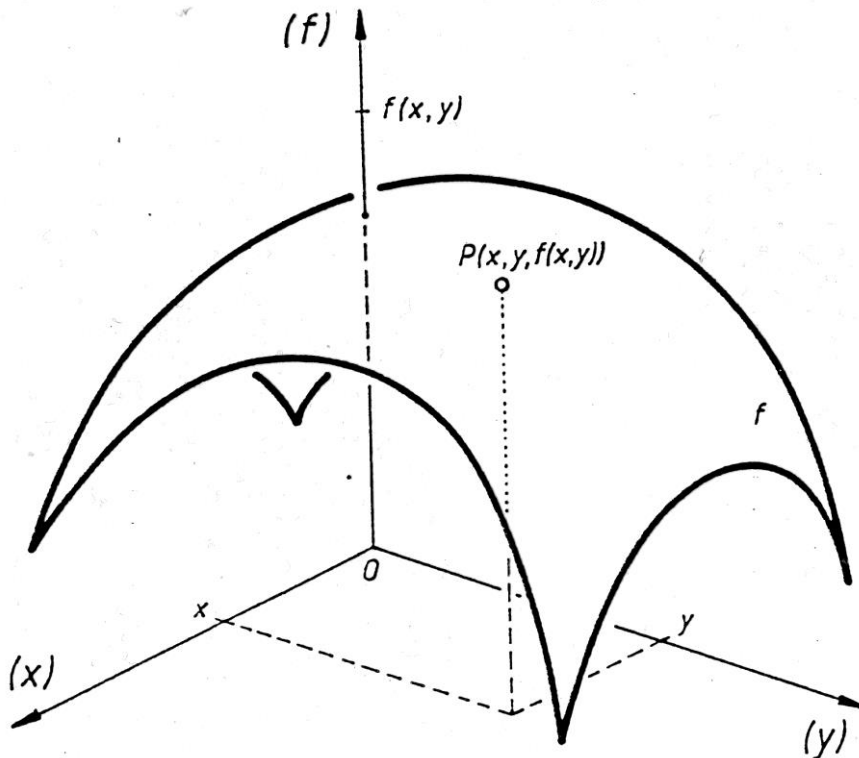


## 2.3 Multivariate functions

$$y = f(\underline{x}) = f(x_1, \dots, x_n): Q \subset \mathbb{R}^n \rightarrow \mathbb{R}$$

Definite mapping of  $Q$  into  $\mathbb{R}$



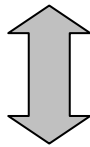
### 2.3.1 Examples, Continuity

- Yield function
- Graph of the function is displayable as curved surface for  $n=2$
- Paraboloid  $y = x_1^2 + x_2^2$
- Objective function in the fodder-mix modell:  
minimal costs:  $60x_1 + 45x_2 + 36x_3 = f(x_1, x_2, x_3)$   
is a linear function

## Remarks concerning the continuity of functions

$n = 1$

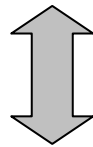
$f(x)$  continuous in  $x_0$ : Is the distance of  $x$  and  $x_0$  small, the distance of  $f(x)$  and  $f(x_0)$  is small as well.



$$\lim_{x \rightarrow x_0} f(x) = f(x_0)$$

$n > 1$   $\underline{x} = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$        $\underline{x}^0 = (x_1^0, x_2^0, \dots, x_n^0) \in \mathbb{R}^n$

$f(\underline{x})$  continuous in  $\underline{x}^0$ : Is the distance of  $\underline{x}$  and  $\underline{x}^0$  small, the distance of  $f(\underline{x})$  and  $f(\underline{x}^0)$  is small as well.



$$\lim_{\underline{x} \rightarrow \underline{x}^0} f(\underline{x}) = f(\underline{x}^0)$$

Concerning this, the distance of n-tuples is defined using the Euclidean norm.

$$\| \underline{x} - \underline{x}^0 \| := \sqrt{\sum_{i=1}^n (x_i - x_i^0)^2}.$$

Heuristics:  $n = 1$       Drawing of the curve without any break  
 $n = 2$       „Curved surface without any hole or crack“.

## 2.3.2 Partial derivative

Gradient of the function in direction of the axes

Partial derivative:

Derivative of the function  $f(\underline{x}) = f(x_1, x_2, \dots, x_n)$  with respect to **one of those variables in each case**, whereby the other variables are concerned as constants.

Also the definition of the partial derivative is done at first at the local point  $\underline{x}^0 \in \mathbb{R}^n$  by using a limit:

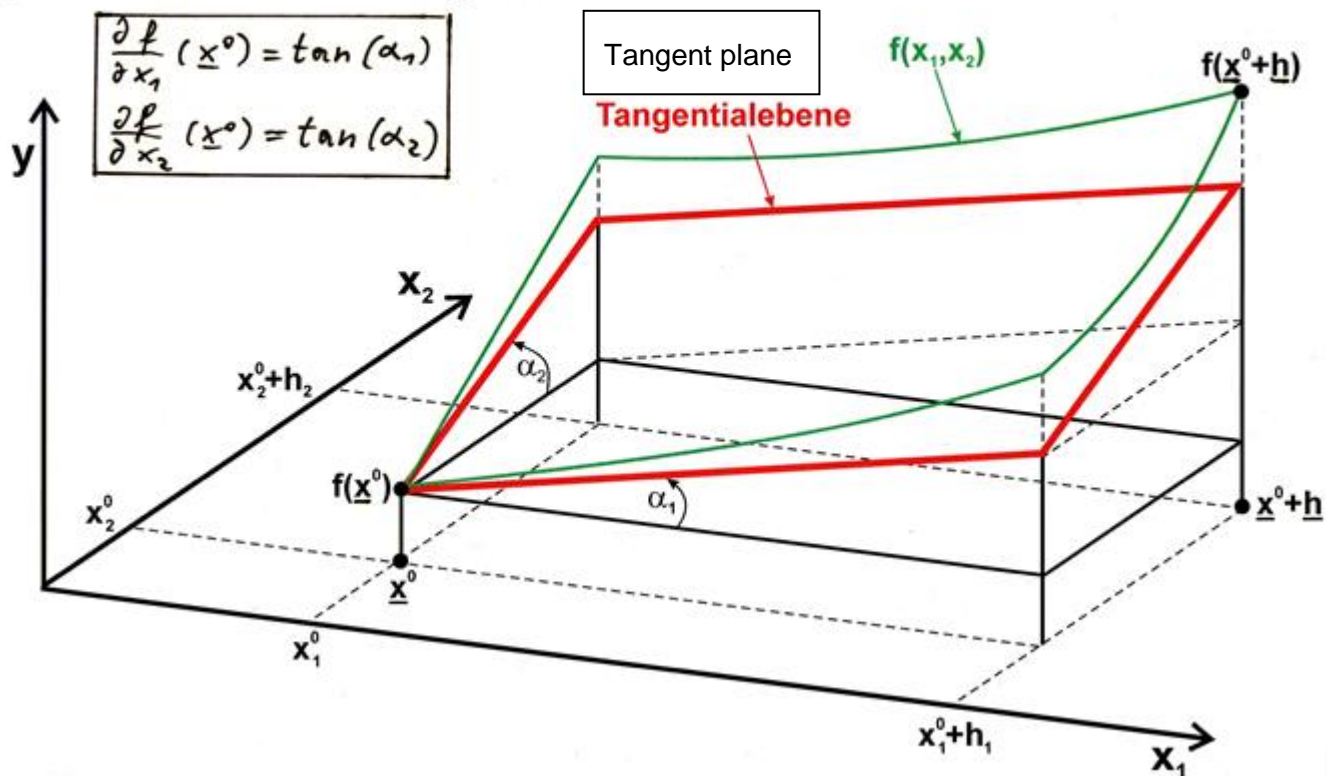
$$\lim_{h \rightarrow 0} \frac{f(x_1^0, \dots, x_i^0 + h, \dots, x_n^0) - f(x_1^0, \dots, x_i^0, \dots, x_n^0)}{h} \quad \text{for } i = 1, \dots, n.$$

Designation:  $\frac{\partial f}{\partial x_i}$  (also  $\frac{\partial}{\partial x_i} f$ ,  $\frac{\partial y}{\partial x_i}$  respectively  $f_{x_i}$ ,  $f_i$ )

Geometrical interpretation:

Gradient of the function in direction of the axes.

# Geometrische Interpretation der partiellen Ableitungen



## 2. Partial derivatives (Second order partial derivatives):

$$\frac{\partial^2 f}{\partial x_i \partial x_j} := \frac{\partial}{\partial x_j} \left( \frac{\partial f}{\partial x_i} \right) \text{ as abbreviated form also } f_{ij};$$

This means that the partial derivative of the function with respect to  $x_i$ , is partially derivated again with respect to the variable  $x_j$ .

Altogether there are  $n^2$  second order derivative functions:

$$\begin{array}{ccc} \frac{\partial^2 f}{\partial x_1^2}, & \frac{\partial^2 f}{\partial x_1 \partial x_2}, & \dots, \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1}, & \frac{\partial^2 f}{\partial x_2^2}, & \dots, \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \vdots & \vdots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1}, & \frac{\partial^2 f}{\partial x_n \partial x_2}, & \dots, \frac{\partial^2 f}{\partial x_n^2} \end{array}$$

Example 1:  $y = f(x_1, x_2) = x_1^2 + x_2^2$

$$\frac{\partial f}{\partial x_1} = 2x_1 \quad , \quad \frac{\partial f}{\partial x_2} = 2x_2$$

Example 2:  $f(x_1, x_2) = 5 + 2x_1 + 5x_1^2 + 8x_1 x_2 + 7x_2 + 5x_2^2$

$$\frac{\partial f}{\partial x_1} = 2 + 10x_1 + 8x_2 \quad , \quad \frac{\partial f}{\partial x_2} = 8x_1 + 7 + 10x_2,$$

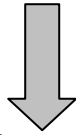
$$\frac{\partial^2 f}{\partial x_1 \partial x_1} = 10 \quad \frac{\partial^2 f}{\partial x_1 \partial x_2} = 8$$

$$\frac{\partial^2 f}{\partial x_2 \partial x_1} = 8 \quad \frac{\partial^2 f}{\partial x_2 \partial x_2} = 10$$

### 2.3.3 Local extrema of multivariate functions

(1) One variable:  $x \in \mathbb{R}$ ,  $y = f(x)$ ,

we are looking for a local maximum respectively minimum of the function  $f(x)$



necessary condition:  $f'(x) = 0$ ,

sufficient condition:  $f''(x) < 0$  local maximum

bzw.  $f''(x) > 0$  local minimum.

(2) n variables:  $(x_1, x_2, \dots, x_n) = \underline{x} \in \mathbb{R}^n$

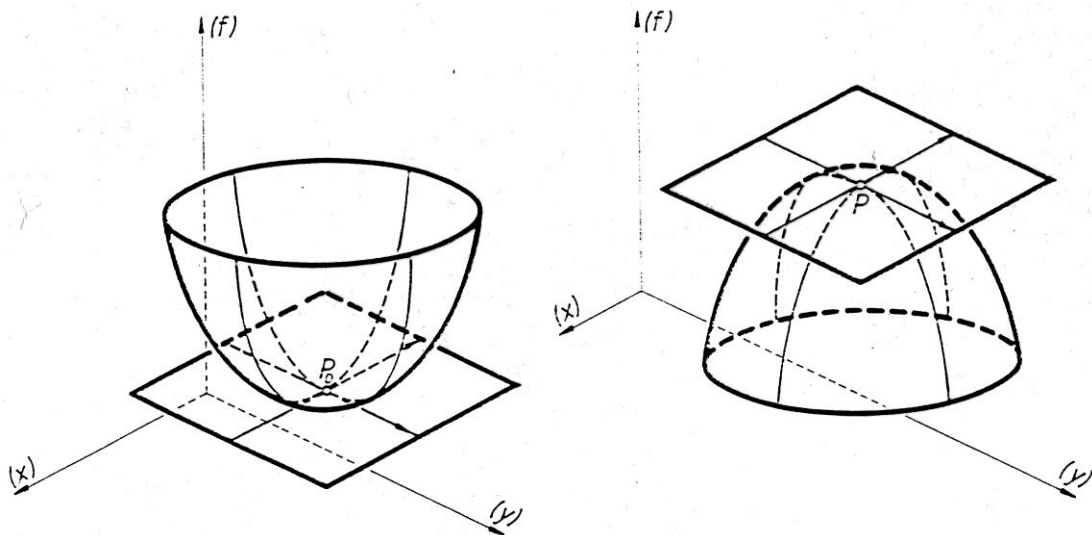
Definition:

The function  $f(x)$  has a local maximum( respectively minimum) at an inner point  $\underline{x}^0$  of the definition range if a real number  $\delta > 0$  exists with  $f(\underline{x}) \leq f(\underline{x}^0)$  (respectively  $f(\underline{x}) \geq f(\underline{x}^0)$ ) for every  $\underline{x} \in \mathbb{R}^n$  with  $\| \underline{x} - \underline{x}^0 \| < \delta$ .

Collective term: local extremum

$\{ \underline{x} \in \mathbb{R}^n \mid \| \underline{x} - \underline{x}^0 \| < \delta \}$  is called an open sphere around  $\underline{x}^0 \in \mathbb{R}^n$

## Local maximum/ Local Minimum





## Necessary condition for relative extrema:

Theorem: If  $f(\underline{x})$  has a local extremum at point  $\underline{x}^0 \in \mathbb{R}^n$  and if all partial derivatives  $\frac{\partial f}{\partial x_i}$  exist at this point, then  $\frac{\partial f}{\partial x_i}(\underline{x}^0) = 0 \quad \forall i = 1, \dots, n.$

We consider  $n$  equations. The solutions of this system of equations are called stationary points.

Furthermore let  $n = 2$ , that means we have two variables  $x_1, x_2$ .

So there is a geometrical interpretation of the necessary conditions:

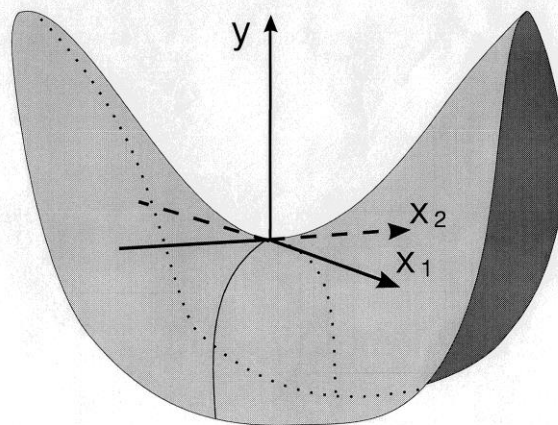
the tangent plane of  $f(x_1, x_2)$  at point  $\underline{x}^0 = (x_1^0, x_2^0)$  is parallel to the  $x_1, x_2$ -plane.

In this case, we also need a sufficient condition for the local extremum.

Example: Saddle surface  $f(x_1, x_2) = x_1 \cdot x_2$

Sattelfläche  $f(x_1, x_2) = x_1 \cdot x_2$

Saddle surface



### Sufficient condition for a relative extremum:

Let  $f(x_1, x_2)$  be defined and continuous in a neighbourhood of a stationary point  $\underline{x}^0 = (x_1^0, x_2^0)$  and let us suppose that all first and second partial derivatives exist and are continuous as well.

Theorem:  $f(x_1, x_2)$  has a local minimum at the stationary point

$$\underline{x}^0 = (x_1^0, x_2^0) \text{ if}$$

$$(1) \quad D = \frac{\partial^2 f}{\partial x_1 \partial x_1}(\underline{x}^0) \cdot \frac{\partial^2 f}{\partial x_2 \partial x_2}(\underline{x}^0) - \left[ \frac{\partial^2 f}{\partial x_1 \partial x_2}(\underline{x}^0) \right]^2 > 0$$

and

$$(2) \quad \frac{\partial^2 f}{\partial x_1 \partial x_1}(\underline{x}^0) > 0.$$

If  $D > 0$  and  $\frac{\partial^2 f}{\partial x_1 \partial x_1}(\underline{x}^0) < 0$ , then  $f$  has a local maximum at  $\underline{x}^0$ .

In Example 2 from section 2.3.2 we defined all first and second partial derivatives for the function

$$f(x_1, x_2) = 5 + 2x_1 + 5x_1^2 + 8x_1 x_2 + 7x_2 + 5x_2^2 .$$

The necessary condition is the linear equality system

$$2 + 10x_1 + 8x_2 = 0$$

$$7 + 8x_1 + 10x_2 = 0 .$$

The solution is the stationary point  $\underline{x}^0 = (1, -\frac{3}{2})$ .

The sufficient condition is

$$(1) \quad \mathbf{D} = 10 \cdot 10 - 8^2 = 36 > 0 \text{ and}$$

$$(2) \quad \frac{\partial^2 f}{\partial x_1 \partial x_1} = 10 > 0 .$$

(1) makes sure there is a local extremum at point  $\underline{x}^0$ ,

(2) shows that the local extremum is a local minimum.