# $TUM\ ModSim,\ SoSe\ 2023$

Mitschriften basierend auf der Vorlesung von Dr. Hans-Joachim Bungartz

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# Introduction

#### About

Hier sind die wichtigsten Konzepte der ModSim Vorlesung von Dr. Hans-Joachim Bungartz im Sommersemester 2023 zusammengefasst.

Die Mitschriften selbst sind in Markdown geschrieben und werden mithilfe einer GitHub-Action nach jedem Push mithilfe von Pandoc zu einem PDF konvertiert.t

Eine stets aktuelle Version der PDFs kann über modsim\_SS23\_IN2010\_merge.pdf heruntergeladen werden.

# Implementation

Außerdem befindet sich eine Implementation von verschiedenen Algorithmen im Ordner /algorithms auf GitHub. Diese sind in Python und unter der Verwendung von NumPy geschrieben.

# How to Contribute

- 1. Fork this Repository
- 2. Commit and push your changes to your forked repository
- 3. Open a Pull Request to this repository
- 4. Wait until the changes are merged

## Contributors

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# Focus Analysis / Calculus

# **Foundations**

#### Functions and their representations

• One-Dimensional

$$f: D \subseteq \mathbb{R}^n \to \mathbb{R}^m, x \mapsto f(x)$$

• Multidimensional

$$f: D \subseteq \mathbb{R} \to \mathbb{R}, x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mapsto f(x) = \begin{pmatrix} f_1(x_1, \dots, x_n) \\ \vdots \\ f_m(x_1, \dots, x_n) \end{pmatrix}$$

## Names for special types of functions

- Curves: n = 1 and  $m \in \mathbb{N}$ 
  - plane curves (2D): n = 1 and m = 2
  - space curves (3D): n = 1 and m = 3
- Surfaces: n=2 and m=3
- Scalar fields:  $n \in \mathbb{N}$  and m = 1
- Vector fields: n = m

#### Topology concepts in higher dimensions

There is an analogous concept to open and closed intervals in multi-dimensional spaces. Given a domain  $D \subset \mathbb{R}^n$  and its complement  $D^c = \mathbb{R}^n \setminus D$ 

- A point x is called *inner point* if there exists an arbitrarily small ball around this points that fullly lies inside D.
- The set of all inner points of D is called the *interior* of D and is denoted as  $\mathring{D}$ .
- The domain is called open if D = D
- A point  $x_0 \in \mathbb{R}^n$  is called *boundary point* if any arbitrarily small ball around this point intersects with both D and its complement  $D^c$
- The set of all boundary points of D is called the boundary of D, denoted  $\partial D$
- The set  $\overline{D} = D \cup \partial D$  is called the *closure* of D

Using these definitions there are multiple attributes assignable to domains.

A domain D is called:

- closed if  $\partial D \subseteq D$ , i.e.  $\bar{D} = D$
- bounded if  $\exists K \in \mathbb{R} : ||x|| < K, \forall x \in D$
- compact if it is closed and bounded
- convex if all points on a straight line between to points in D are themselves element of D

## Continuity

We define continuity in multi-dimensional spaces using converging vector sequences. A sequence  $(x^{(k)})$  converges to the limit x if

$$\lim_{k \to \infty} ||x^{(k)} - x|| = 0$$

Converges of a vector sequence is also equivalent to the convergence of all components.

A vector function is then called continuous at  $a \in D$  if for all sequences  $(x^{(k)})_{k \in \mathbb{N}_0}$  in D converging to a the corresponding sequence  $(f(x^{(k)}))_{k \in \mathbb{N}_0}$  in  $\mathbb{R}^m$  converges to f(a) and continuous on D if this holds for all points  $a \in D$ 

# Partial Differentiation

#### Gradient

The Gradient of a function gives the direction of the steepest ascent of the function. It requires that f represents a scalar field.

When applying the limit definition of the derivative to a function in higher dimesions it is not clear from which direction the derivative should be taken.

Using

$$\frac{\partial f}{\partial v}(a) = \lim_{h \to 0} \frac{f(a+hv) - f(a)}{h}$$

we can define the directional derivative of a function  $f: \mathbb{R}^n \to \mathbb{R}$  along a vector  $v \in \mathbb{R}^n$  at a point  $a \in \mathbb{R}^n$ .

If we use the coordinate vectors  $e_i$  as basis vectors for  $\mathbb{R}^n$  we can define the *Gradient* of f at a as

$$\nabla f(a) = \operatorname{grad} f(a) = \begin{pmatrix} \frac{\partial f}{\partial x_1}(a) \\ \vdots \\ \frac{\partial f}{\partial x_n}(a) \end{pmatrix}$$

For continuous functions the directional derivative at the point a along a vector v can be computed as

$$\frac{\partial f}{\partial v}(a) = \langle \nabla f(a), v \rangle$$

Example:

$$f(x,y) = x^2 + y^2 \rightarrow \nabla f(a) = \begin{pmatrix} 2x \\ 2y \end{pmatrix}$$

#### Hessian Matrix

The Hessian matrix of a function  $f: \mathbb{R}^n \to \mathbb{R}$  at a point  $a \in \mathbb{R}^n$  is the matrix of all second partial derivatives of f at a.

$$H_f(a) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1^2}(a) & \frac{\partial^2 f}{\partial x_1 \partial x_2}(a) & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n}(a) \\ \frac{\partial^2 f}{\partial x_2 \partial x_1}(a) & \frac{\partial^2 f}{\partial x_2^2}(a) & \dots & \frac{\partial^2 f}{\partial x_2 \partial x_n}(a) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1}(a) & \frac{\partial^2 f}{\partial x_n \partial x_2}(a) & \dots & \frac{\partial^2 f}{\partial x_n^2}(a) \end{pmatrix}$$

Example:

$$f(x,y) = x^2 + y^2 \to H_f(a) = \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$$

#### Jacobian Matrix

The Jacobian matrix of a function  $f: \mathbb{R}^n \to \mathbb{R}^m$  at a point  $a \in \mathbb{R}^n$  is the matrix of all partial derivatives of f at a.

In contrast to the Hessian matrix and the gradient the Jacobian matrix works for vector fields.

$$Df(a) = J_f(a) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(a) & \frac{\partial f_1}{\partial x_2}(a) & \dots & \frac{\partial f_1}{\partial x_n}(a) \\ \frac{\partial f_2}{\partial x_1}(a) & \frac{\partial f_2}{\partial x_2}(a) & \dots & \frac{\partial f_2}{\partial x_n}(a) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1}(a) & \frac{\partial f_m}{\partial x_2}(a) & \dots & \frac{\partial f_m}{\partial x_n}(a) \end{pmatrix} = \begin{pmatrix} \nabla f_1(a)^T \\ \nabla f_2(a)^T \\ \ddots \\ \nabla f_m(a)^T \end{pmatrix}$$

Example:

$$f(x,y) = \begin{pmatrix} x^2 + y \\ x^2 + y^2 \end{pmatrix} \rightarrow J_f(a) = \begin{pmatrix} 2x & 1 \\ 2x & 2y \end{pmatrix}$$

#### Calculation rules for the Jacobian

• Addition rule:  $J(f+g) = J_f + J_g$ 

• Homogeneous rule:  $J(cf) = cJ_f$ • Product rule:  $J(f^T \cdot g) = f(x)^T J_g(x) + g(x)^T J_f(x)$ 

# Laplace Operator

The Laplace operator is a second order partial derivative operator. It is defined on Scalar fields and is used to compute the rate of change of a scalar field.

$$\Delta f = \nabla^2 f = \sum_{i=1}^n \frac{\partial^2 f}{\partial x_i^2}$$

Example:

$$f(x,y) = x^2 + y^2 \to \Delta f(a) = 2 + 2 = 4$$

### Divergence

The Divergence of a vector field is the rate of shrinkage or expansion around a point. It is defined as the sum of the partial derivatives of the components of the vector field.

$$\operatorname{div} f = \sum_{i=1}^{n} \frac{\partial f_i}{\partial x_i} = \nabla \cdot f$$

Example:

$$f(x,y) = \begin{pmatrix} x^2 \\ y^2 \end{pmatrix} \to \operatorname{div} f(a) = \frac{\partial f_1}{\partial x} + \frac{\partial f_2}{\partial y} = 2x + 2y$$

#### Curl / Rotation

The Curl of a vector field is the rate of rotation around a point. It is defined as the cross product of the partial derivatives of the components of the vector field.

$$\operatorname{rot} f = \nabla \times f = \begin{pmatrix} \frac{\partial f_3}{\partial x_2} - \frac{\partial f_2}{\partial x_3} \\ \frac{\partial f_1}{\partial x_3} - \frac{\partial f_3}{\partial x_1} \\ \frac{\partial f_2}{\partial x_1} - \frac{\partial f_1}{\partial x_2} \end{pmatrix}$$

Example:

$$f(x,y) = \begin{pmatrix} x^2y \\ y^2x \\ yz \end{pmatrix} \to \text{rot} f(a) = \begin{pmatrix} z \\ 0 \\ y^2 - x^2 \end{pmatrix}$$