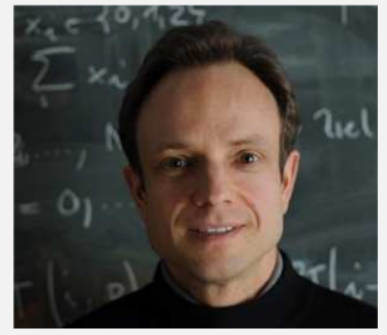


# Probevorlesung Maturandentage Formerhaltende Interpolation

Prof. R. Hiptmair, SAM, ETH Zurich

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Aufgabe :

Modelliere die Beziehung physikalischer Größen  $t, y$  durch eine (einfache) Funktion

$$f: D \subset \mathbb{R} \longrightarrow \mathbb{R}$$

gestützt auf (genaue) Messwerte

$$(t_i, y_i) \in \mathbb{R} \times \mathbb{R}, t_i \in D, t_i < t_{i+1}, i=0, \dots, n$$

$$n \in \mathbb{N}.$$

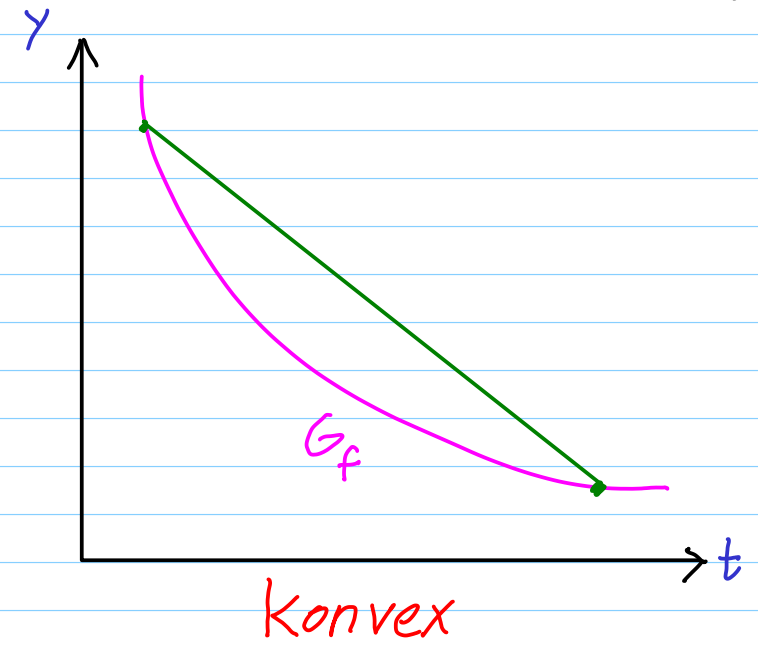
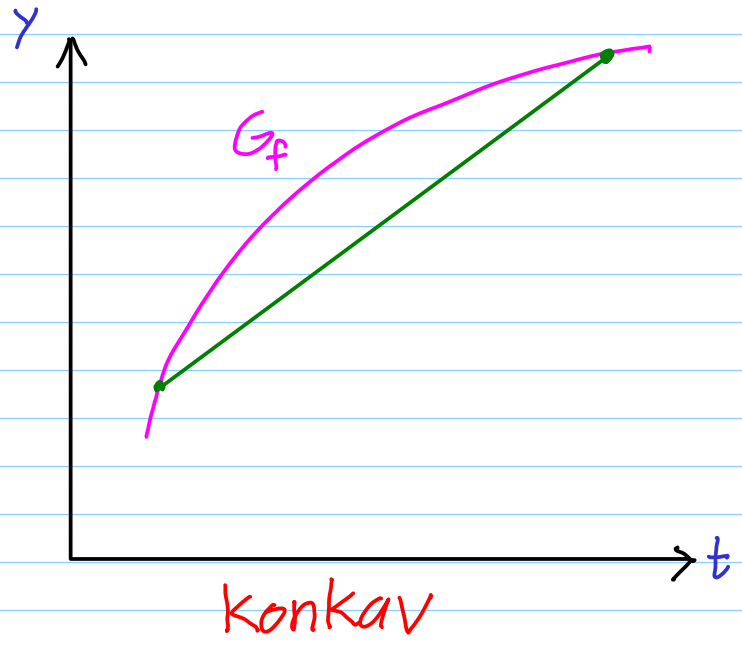
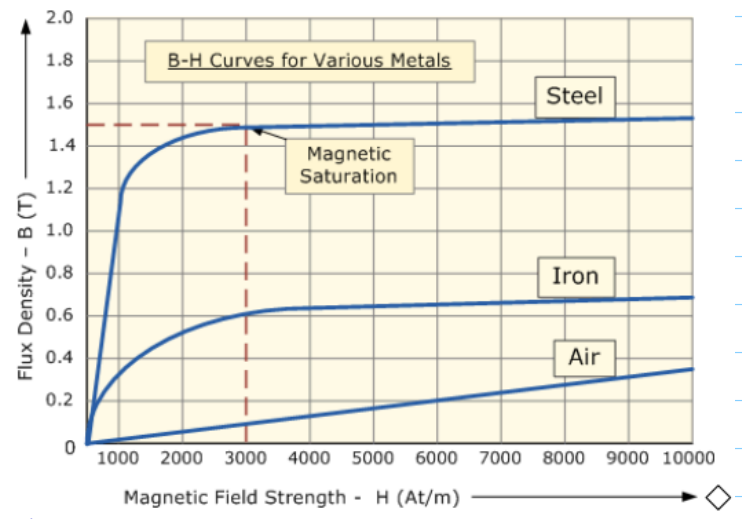
▷ Interpolationsbedingungen

$$f(t_i) = y_i, \quad i = 0, \dots, n$$

Example 1.4.1 (Magnetization curves).

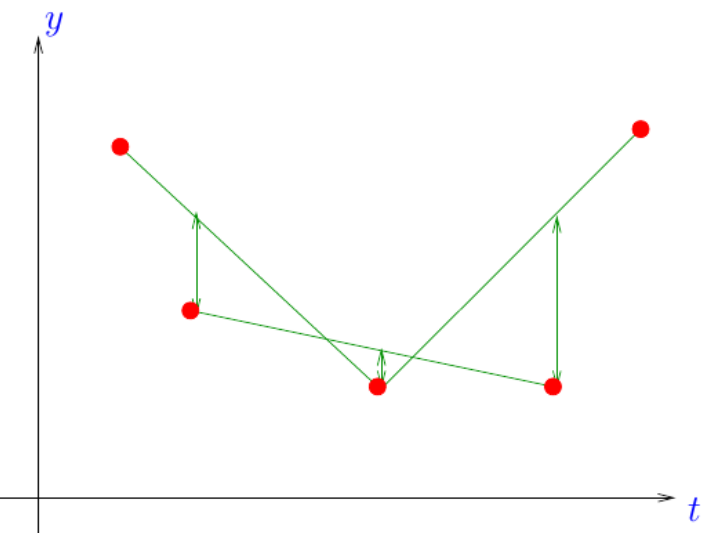
For many materials physics stipulates properties of the functional dependence of magnetic flux  $B$  from magnetic field strength  $H$ :

- $H \mapsto B(H)$  monotone (increasing),
- $H \mapsto B(H)$  concave
- $H \mapsto B(H)$  smooth



# Formeigenschaften von Daten:

**Definition 1.4.2** (monotonic data).  
 The data  $(t_i, y_i)$  are called **monotonic**, when  $y_i \geq y_{i-1}$  or  $y_i \leq y_{i-1}, i = 1, \dots, n$ .



**Definition 1.4.3** (Convex/concave data).  
 The data  $\{(t_i, y_i)\}_{i=0}^n$  are called **convex** (**concave**) if

$$\Delta_j \stackrel{(\geq)}{\leq} \Delta_{j+1}, \quad j = 1, \dots, n-1,$$

$$\Delta_j := \frac{y_j - y_{j-1}}{t_j - t_{j-1}}, \quad j = 1, \dots, n.$$

"connecting lines above data points"

Zurück zur Interpolation:

Daten  $(t_i, y_i)_{i=0}^n \longrightarrow$  Funktion  $f: D \subset \mathbb{R} \rightarrow \mathbb{R}$   
 $f(t_i) = y_i$

Goal: **shape preserving interpolation:**

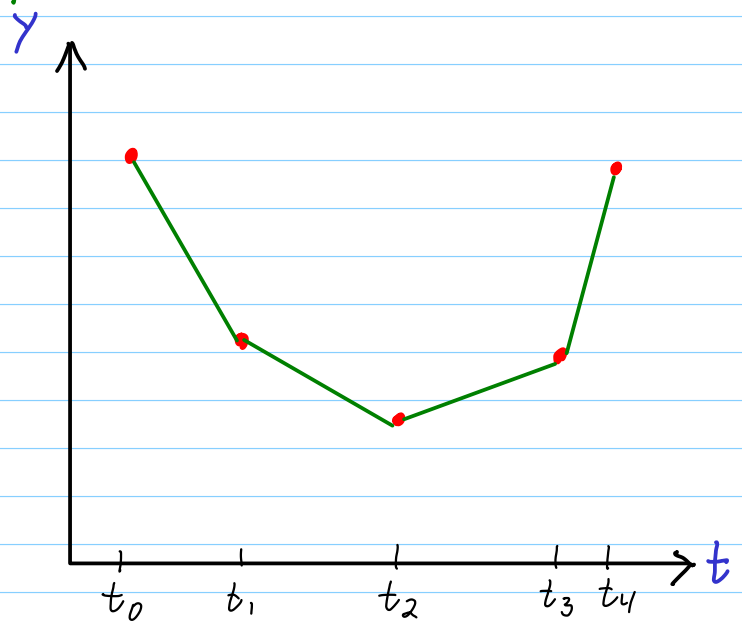
positive data	$\longrightarrow$	positive interpolant $f$ ,
monotonic data	$\longrightarrow$	monotonic interpolant $f$ ,
convex data	$\longrightarrow$	convex interpolant $f$ .

# Noch ambitionierter: Lokale Formerhaltung

$I \subset D$ :

positive data in $I$	$\longrightarrow$	locally positive interpolant $f _I$ ,
monotonic data in $I$	$\longrightarrow$	locally monotonic interpolant $f _I$ ,
convex data in $I$	$\longrightarrow$	locally convex interpolant $f _I$ .

Option: Stückweise lineare Interpolation



$\bullet \leftrightarrow (t_i, y_i)$

$\text{---} \stackrel{!}{=} \text{Streckenzug / Polygon}$

**Theorem 1.4.4** (Local shape preservation by piecewise linear interpolation).  
 Let  $s \in C([t_0, t_n])$  be the piecewise linear interpolant of  $(t_i, y_i) \in \mathbb{R}^2, i = 0, \dots, n$ , for every subinterval  $I = [t_j, t_k] \subset [t_0, t_n]$ :

if $(t_i, y_i) _I$ are positive/negative	$\Rightarrow$	$s _I$ is positive/negative,
if $(t_i, y_i) _I$ are monotonic (increasing/decreasing)	$\Rightarrow$	$s _I$ is monotonic (increasing/decreasing),
if $(t_i, y_i) _I$ are convex/concave	$\Rightarrow$	$s _I$ is convex/concave.



Einzigster Makel: Fehlende Glattheit

Option: Stückweise kubische Interpolation

$s := f|_{[t_{i-1}, t_i]}$  ist ein Polynom vom Grad 3:

$$s(t) = f|_{[t_{i-1}, t_i]}(t) = a_i t^3 + b_i t^2 + c_i t + d_i, \quad a_i, b_i, c_i, d_i \in \mathbb{R}$$

$\downarrow$                      $\downarrow$                      $\downarrow$                      $\downarrow$   
 4 "Freiheitsgrade"

Given: data values  $y_{i-1}, y_i \in \mathbb{R}$ , slopes  $c_{i-1}, c_i \in \mathbb{R}$

> There is a unique polynomial  $s$  of degree  $\leq 3$  with

$$s(t_{i-1}) = y_{i-1} \quad , \quad s(t_i) = y_i \quad , \quad s'(t_{i-1}) = c_{i-1} \quad , \quad s'(t_i) = c_i .$$

Interpolationsbedingungen                    Fixierte Steigungen

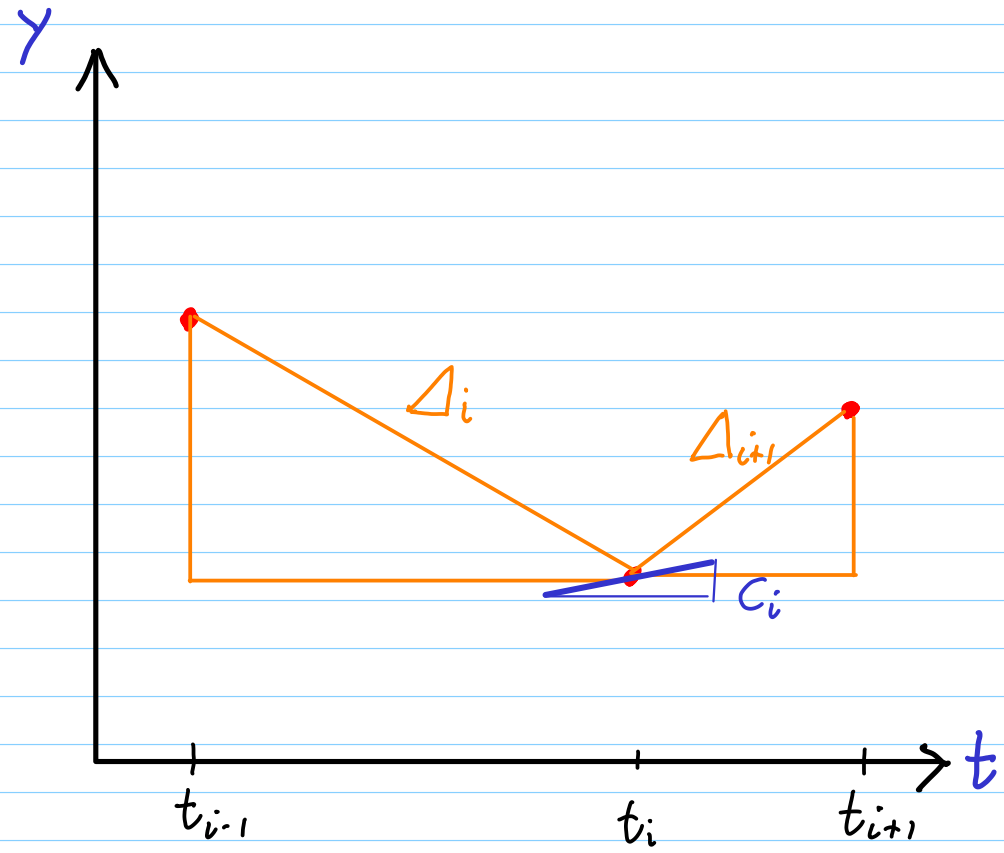
▷ "Automatisch"  $f \in C^1$  (stetig differenzierbar)

Wie wählt man die Steigungen  $c_i$ ?

Naheliegend: gewichtetes Mittel

$$c_i = \begin{cases} \Delta_1 & , \text{ for } i = 0, \\ \Delta_n & , \text{ for } i = n, \\ \frac{t_{i+1}-t_i}{t_{i+1}-t_{i-1}} \Delta_i + \frac{t_i-t_{i-1}}{t_{i+1}-t_{i-1}} \Delta_{i+1} & , \text{ if } 1 \leq i < n. \end{cases} \quad \Delta_j := \frac{y_j - y_{j-1}}{t_j - t_{j-1}}, \quad j = 1, \dots, n.$$

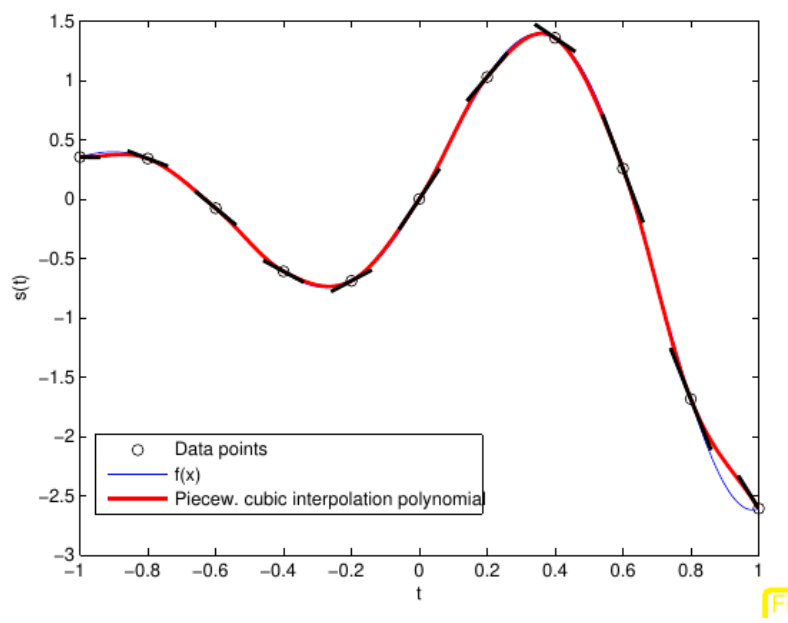
(1.4.6)



Example 1.4.7 (Piecewise cubic Hermite interpolation with averaged local slopes).

Data points:

- 11 equispaced nodes
- $t_j = -1 + 0.2j, \quad j = 0, \dots, 10.$
- in the interval  $I = [-1, 1],$
- $y_i = f(t_i)$  with
- $f(x) := \sin(5x) e^x .$



Use of weighted averages of slopes as in (1.4.6).



Idee: Begrenzung der Steigung

$$c_i = \begin{cases} 0, & \text{wenn } \text{sgn}(\Delta_i) \neq \text{sgn}(\Delta_{i+1}), \\ \text{"geeignetes Mittel" von } \Delta_i, \Delta_{i+1} & \text{sonst.} \end{cases}$$

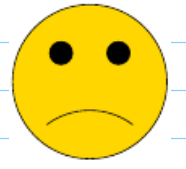
$\approx 0$ , falls  $\Delta_i \approx 0$  oder  $\Delta_{i+1} \approx 0$

▷ Harmonisches Mittel

$$h(a,b) = \frac{2}{\frac{1}{a} + \frac{1}{b}}$$

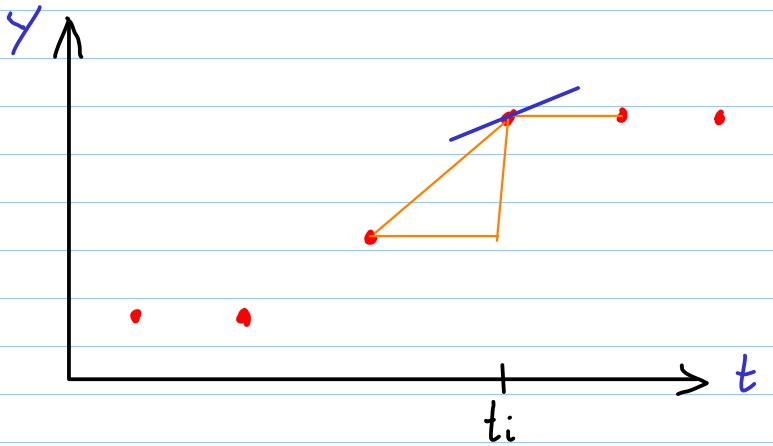
Special choice: weighted harmonic mean of local slopes

$$c_i = \frac{3(h_{i+1} + h_i)}{\frac{2h_{i+1} + h_i}{\Delta_i} + \frac{2h_i + h_{i+1}}{\Delta_{i+1}}}, \quad i = 1, \dots, n-1, \quad h_i := t_i - t_{i-1}. \quad (1.4.9)$$



Keine lokale Monotonieerhaltung

Problem "Überschiessen"



Mittelung:

$$c_i > 0 !$$

**Theorem 1.4.10** (Monotonicity preservation of limited cubic Hermite interpolation).  
 The cubic Hermite interpolation polynomial with slopes as in (1.4.9) provides a local monotonicity preserving  $C^1$ -interpolant.

