

**TODA** A Kurven

Globalübung OG  
Mathe 3 WS20

06.01.2021

①

Lambert Theisen

N: PM, IPM, Shift, Rayleigh-Coeff.

② Klausurammlung, Piz Testat Mode!

A

Kurven:  $\rightarrow$  stetige Abbildung  $\gamma: [a, b] \rightarrow \mathbb{R}^n$ .

• Weg:  $\hat{=} 3D$   $\Gamma = \gamma([a, b])$   $\xrightarrow{\text{L}} \text{Parametrisierung}$  von  $\Gamma$

• Einfachheit: Falls  $\gamma$  injektiv ("doppel punktfrei")  $\xrightarrow{\text{L}} \text{nicht einfach}$

• Geschlossenheit: Falls  $\gamma(a) = \gamma(b)$   $\xrightarrow{\text{L}} \text{18} \checkmark$

• Einfachheit  $\leftarrow$  Falls  $\gamma$  geschlossen &  $\gamma|_{[a, b]}$  injektiv  $\xrightarrow{\text{L}} \text{18} \checkmark$

• C<sup>1</sup>-Kurve: Falls  $\gamma$  stetig diff'bar  $\xrightarrow{\text{L}} \checkmark$

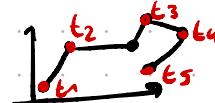
• Tangentialvektor im Plat.  $\gamma(t)$ :  $\gamma'(t) = (\gamma_1'(t), \dots, \gamma_n'(t))^T$

• Glättigkeit: Falls  $\gamma$  eine  $C^1$ -kurve &  $\|\gamma'(t)\| \neq 0 \forall t \in [a, b]$

• Tangentialeinheitsvektor für glatte Kurven:  $T: [a, b] \rightarrow \mathbb{R}^n$ :  $T(t) := \frac{\gamma'(t)}{\|\gamma'(t)\|} \Rightarrow \|T(t)\| = 1 \checkmark$

• Umparametrisierung:  $\tilde{\gamma} := \gamma \circ \phi: [c, d] \rightarrow [a, b] \rightarrow \mathbb{R}^n$ ,  $t \mapsto \gamma(\phi(t))$   $\xrightarrow{\text{Skalar!}} \checkmark$   
 geht nur für glatte Kurven!  
 aber  $\tilde{\gamma}' \neq 1$ !

• Stückweise Eigenschaften:



Wenn  $\tilde{t} = \{\tilde{t}_1, \dots, \tilde{t}_n\}$  & Menge endlich

$|\tilde{t}| = 5$  endlich

$\rightarrow$  z.B. Polygonzug stückweise glatt

• Bogenlänge für  $C^1$ -Kurven:

$$\rightarrow L(\gamma) = \int_a^b \|\gamma'(t)\| dt$$

"Quasi Maßband an Kurve anlegen"

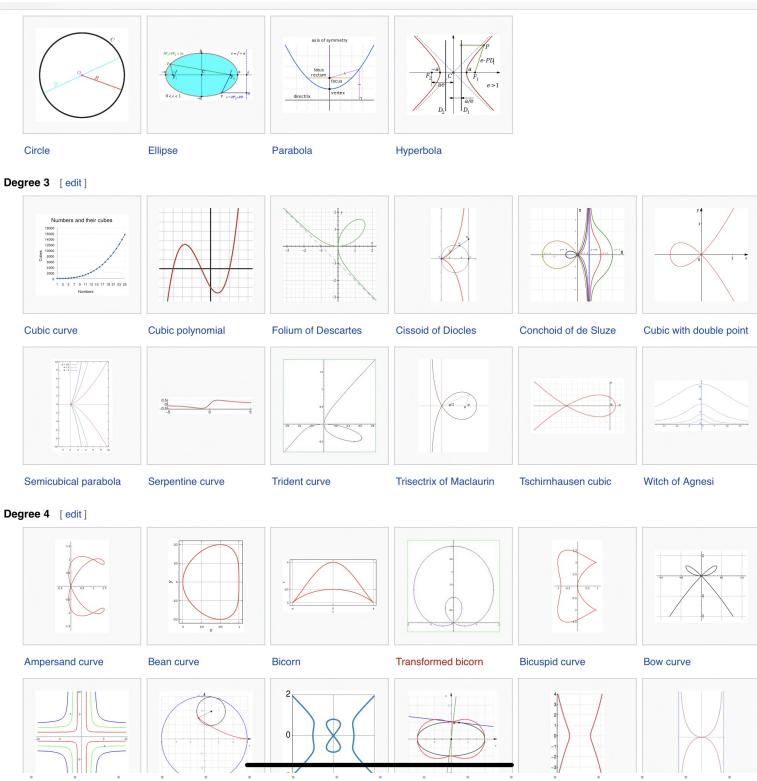
• Invarianz Bogenlänge unter Umparametrisierung:

Sei  $\phi: [c, d] \rightarrow [a, b]$  Diffeomorphismus

und  $\tilde{\gamma} := \gamma \circ \phi$ , dann

$$L(\tilde{\gamma}) = \int_c^d \|\tilde{\gamma}'(s)\| ds = \int_a^b \|\gamma'\| dt = L(\gamma)$$

$\|\gamma'\| = 1$  wäre schön... 😐

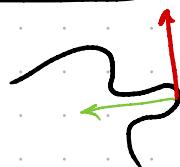


Natürliche Parametrisierung:  $\gamma: [0, L] \rightarrow \mathbb{R}^n$  ist nat. Param. eines Weges  $\Gamma$  falls  $\|\dot{\gamma}(t)\| = 1$ . [Einfaches Ablesen der Bogenlänge]

→ Konstruktion: Benutze Diffeomorphismus  $\psi: [a, b] \rightarrow [0, L]$

$$\text{mit } t \mapsto \psi(t) := \int_a^t \|\dot{\gamma}(s)\| ds \quad \begin{aligned} \gamma: [a, b] &\rightarrow \mathbb{R} \\ \psi(t) &= \int_a^t \sqrt{\dot{\gamma}(s)} ds = \frac{1}{2} t^2 - 2a \end{aligned}$$

Normaleneinheitsvektor: Sei  $T(t) := \frac{\dot{\gamma}(t)}{\|\dot{\gamma}(t)\|}$  Tang'enh'ktor, dann

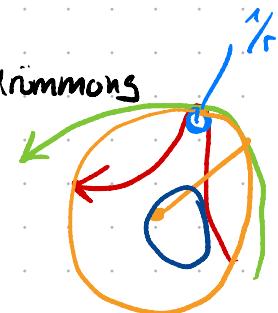


$$N(t) := \frac{T'(t)}{\|T'(t)\|} \quad \text{Norm'enh'ktor.}$$

Krümmung: Sei  $\gamma$  eine natürliche Param., dann ist die Krümmung von  $\gamma$  in  $\gamma$ :

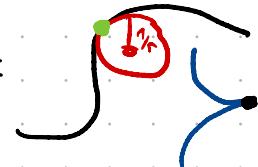
$$\kappa(\gamma) := \|T'(\gamma)\|$$

Kreiskrümmung:  $\gamma: [0, 2\pi] \rightarrow \mathbb{R}$ , stetig  $\gamma(s) = \begin{bmatrix} r \cos(\frac{s}{r}) \\ r \sin(\frac{s}{r}) \end{bmatrix}$



$$\dot{\gamma}(s) = \begin{bmatrix} -\sin(\frac{s}{r}) \\ \cos(\frac{s}{r}) \end{bmatrix} \Rightarrow T(s) = \frac{\dot{\gamma}(s)}{\|\dot{\gamma}(s)\|_2} = \dot{\gamma}(s) \text{ weil } \|\dot{\gamma}(s)\|_2 = \sqrt{\sin^2 + \cos^2} = 1$$

$$\Rightarrow T'(s) = \begin{bmatrix} -\frac{1}{r} \cos(\frac{s}{r}) \\ -\frac{1}{r} \sin(\frac{s}{r}) \end{bmatrix} \Rightarrow \kappa(\gamma) = \sqrt{\frac{1}{r^2} (\sin^2 + \cos^2)} = \frac{1}{r}$$



Beispiel: a) Check: Bsp. muss erst parametrisiert werden!

$$x^{2/3} + y^{2/3} = a^{2/3} \text{ mit } \gamma: [0, 2\pi] \rightarrow \mathbb{R}^2$$

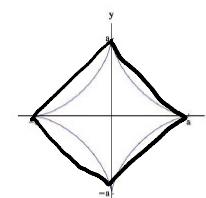
$$[a \cos^3(t)]^{2/3} + [a \sin^3(t)]^{2/3} = a^{2/3}$$

$$\Leftrightarrow a^{2/3} \cos^2(t) + a^{2/3} \sin^2(t) = a^{2/3} \quad \checkmark$$

b) Bogenlänge:  $L(\gamma) = \int_0^{2\pi} \|\dot{\gamma}(t)\|_2 dt$

Daher  $\dot{\gamma}(t) = \begin{bmatrix} \frac{d}{dt}(a \cos^3(t)) \\ \frac{d}{dt}(a \sin^3(t)) \end{bmatrix} = \begin{bmatrix} -3a \cos^2(t) \sin(t) \\ 3a \sin^2(t) \cos(t) \end{bmatrix}$

Aufgabe 68. (Bogenlänge einer Kurve)



The so-called astroid (Sternkurve) is described by the Cartesian equation

$$x^{2/3} + y^{2/3} = a^{2/3}$$

for some  $a > 0$ .

- a) Verify that the astroid can be expressed in parametric form  $\gamma: [0, 2\pi] \rightarrow \mathbb{R}^2$  for

$$x(t) = a \cos^3 t \quad ; \quad y(t) = a \sin^3 t$$

- b) Find the length of the astroid.

- c) Find the area of the astroid using the change of variables  $F: (x, y) \mapsto (r, \varphi)$  defined as

$$x = r \cos^3 \varphi \quad , \quad y = r \sin^3 \varphi$$

Begin by verifying that

$$J(r, \varphi) = \begin{vmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \varphi} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \varphi} \end{vmatrix} = \frac{8}{3} r (1 - \cos 4\varphi)$$

and notice that we can describe the astroid in terms of the coordinates  $(r, \varphi)$  as  $0 \leq r \leq a$  and  $0 \leq \varphi \leq 2\pi$ .

$$\Rightarrow \|\delta'(t)\|_2 \stackrel{a>0}{\geq} 3a \sqrt{(\cos^2 \sin)^2 + (\sin^2 \cos)^2} = 3a \underbrace{\sqrt{(\cos^2 + \sin^2)(\sin^2 \cos^2)}}_{=1}$$

$$\Rightarrow L(\gamma) = \int_0^{2\pi} \|\delta'(t)\| dt = 3a \int_0^{2\pi} \sqrt{\sin^2(t) \cos^2(t)} dt \\ = 3a \int_0^{2\pi} \left( \underbrace{[\sin(t) \cos(t)]^2}_{=\frac{1}{2} \sin(2t)} \right)^{1/2} dt = 3a \int_0^{2\pi} \left| \frac{1}{2} \sin(2t) \right| dt \\ = 12a \int_0^{\pi/2} \frac{1}{2} \sin(2t) dt = 12a \cdot \frac{1}{2}\pi = 6a\pi$$

check!

c) Fläche (nicht so wichtig)  
siehe Tips

$$A = \int_0^a \int_0^{2\pi} 1 \cdot J(r, \varphi) \\ = \int_0^a \int_0^{2\pi} \frac{3}{8}r(1 - \cos(4\varphi)) dr d\varphi \\ = \dots = \frac{3}{8}\pi a^2$$

$$\lambda_1 = \max(\sigma(A))$$

## N Poweriteration (Vektoriteration)

```

• Let  $A \in \mathbb{R}^{n \times n}$ 
• Choose  $x^0 \in \mathbb{R}^n \setminus \{0\}$ 
• While ( $k \leq k_{\max} \wedge |\rho_A(x^k) - \rho_A(x^{k+1})| > \varepsilon$ )
     $\tilde{x}^k = Ax^{k-1}$  // iterate
     $x^k = \tilde{x}^k / \|\tilde{x}^k\|$  // normalize
     $\lambda^k = \rho_A(x^k)$  // eval estimate
• Return  $(\lambda^k, x^k)$ 

```

"Mehrmaliges Anwenden der Matrix"

$$x_1 = Ax_0$$

$$x_2 = A \cdot x_1$$

(Rayleigh Quotient s.b.)  
Aus  $x$  Näherung  $\Rightarrow \lambda$  Näherung

$$q = \left| \frac{\lambda_{k+1}}{\lambda_k} \right|$$

$\rightarrow$  Convergence (usually): I)  $\lim_{k \rightarrow \infty} \lambda^k \rightarrow \lambda_1$  Fehler:  $e \in O(q^k)$

II)  $\lim_{k \rightarrow \infty} x^k \rightarrow x^* \in E(\lambda_1)$   $e \in O(q^k)$

Rayleigh-Quotient:  $[\rho_A(x) =] R(x) := \frac{x^T A x}{x^T x}$  für  $A$  symmetrisch.

$\lambda_1 = \inf_{x \in \mathbb{R}^n \setminus \{0\}} R(x) \leq R(x) \leq \max_{x \in \mathbb{R}^n \setminus \{0\}} R(x) = \lambda_1$  (Rayleigh Prinzip)

Finde Eigenwert zu gegebenem Eigenvektor:

$$R(v_i) = \lambda_j v_i$$

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(Shifted)  
Inverse Power Iteration: ("PM für  $A^{-1}$  findet kleinsten EWert")

One Variant:

Let  $A \in \mathbb{R}^{n \times n}$  and Shift  $\mu$   
 Choose  $x^0 \in \mathbb{R}^n \setminus \{0\}$   
 While ("not finished") do  
     | Solve  $(A - \mu I)x^{k+1} = x^k$  ( $\Leftrightarrow x^{k+1} = (A - \mu I)^{-1}x^k$ )  
     | Normalize  $x^{k+1}$   
     |  $\gamma^{k+1} = \|x^{k+1}\|_2$   
     | Return  $(\gamma^k, x^k)$



→ Shifted IPM konvergiert zu  $\lambda_{\text{closet to } \mu}$

→ Nützlich wenn Näherung bekannt oder für Nicht-extreme EWerte.

Demo:

→ Analysis: Kurven

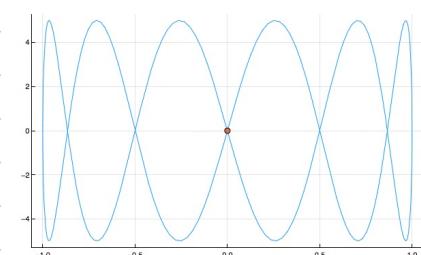
→ Numerik: PM, IPM, SIPM

#### Define a Curve

Define the curve  $\gamma$ :

$$\gamma : [0, 2\pi] \rightarrow \mathbb{R}^2, t \mapsto \gamma(t) = (\sin(t/3), 5 \sin(2t))^T$$

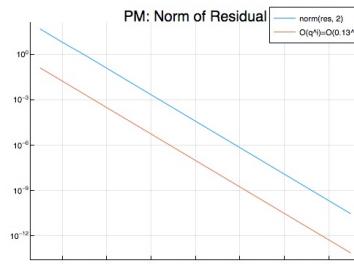
```
v (generic function with 1 method)
• v(t) = [sin(t/3), 5sin(2t)]
• @bind t Slider(0::20:2π::20, show_value=true)
```



```
begin
plot(t->v(t)[1], t->v(t)[2], 0, 6π, leg=false)
scatter!([v(t)[1]], [v(t)[2]])
end
```

#### Residual Error

The normed error of the residual  $e = \|x_k - x^*\|_2$  is in  $\mathcal{O}(q^k)$  with the eigenvalue ratio  $q = \lambda_1/\lambda_2$  of the considered matrix.



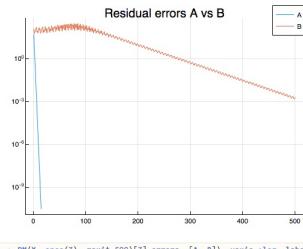
```
plot([pm[3].errors, 1 + abs(eigvals(A)[end-1] / eigvals(A)[end])^i] for i = 1 : size(pm[3].errors, 1)], pm[3].errors, log, title="PM: Norm of Residual", label="norm(res, 2)", "O(q^k)=O($sprintf('k,2^k, abs(eigvals(A)[end-1] / eigvals(A)[end]))^k)"
```

#### Comparison of Matrices with Different Fundamental Ratios

Matrix A has ratio  $q = 0.13333333333333336$

Matrix B has ratio  $q = 0.977623158075516$

Therefore, much faster converge for A.

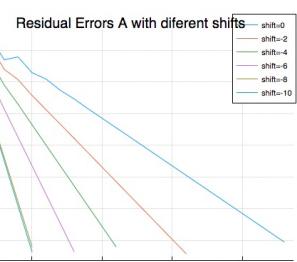


```
plot(snap(X -> PM(X, ones(S)), tol=1E-5, shift=X[3].errors, shifts), yaxis=log, label=["A", "B"], title="Residual errors A vs B")
```

#### Check the Convergence Behavior for Different Shifts

Notice that better shifts significantly improve the performance of the algorithm. Shift eight and nine are the same because they have the same absolute distance to the real eigenvalue.

```
shifts = 0:-2:-10
# Compare zeros shift with good estimation
shifts = 0:-2:-10 # real lowest eval is -9
```



```
plot(snap(X -> PM(X, ones(S)), tol=1E-5, shift=X[3].errors, shifts), yaxis=log, label="Residual errors A with different shifts")
```