

**GraphiCon'98**  
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**Estimating Criteria for Fitting B-spline Curves:  
Application to Data Compression**

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# Outline

## ✉ Survey of the present compression methods

- ↪ Strategies based on polygonal curves
- ↪ Strategy based on spline curves

## ✉ A new compression strategy

- ↪ Data fitting with B-splines
- ↪ Criteria for estimating data approximation
- ↪ Reduction technique
- ↪ Result comparison

## ✉ Conclusion

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## ☒ Conclusion

# Survey of the present methods

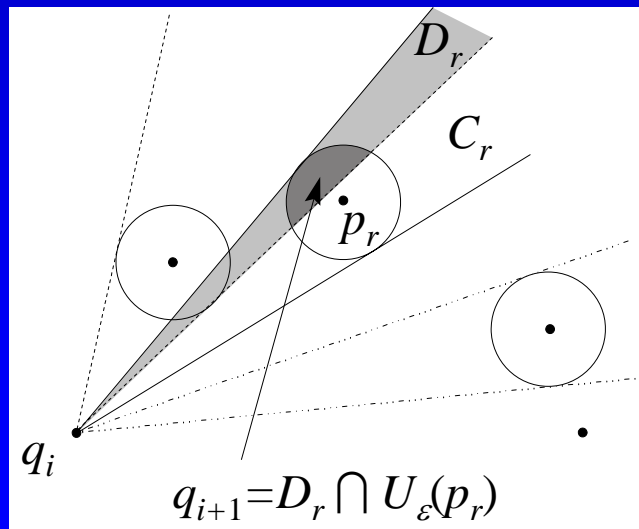
## ↪ Strategies based on polygonal curves

Inputs  $\Rightarrow$  A polygonal curve  $P = (p_0, \dots, p_n)$   
 $\Rightarrow$  A tolerance  $\varepsilon \geq 0$

Output  $\Rightarrow$  A polygonal curve  $Q = (q_0, \dots, q_m)$   
 with  $m < n$  such as  $d(P, Q) \leq \varepsilon$

### ■ The intersecting cones method (E. Arge & M. Daehlen)

General  
 strategy



Intersection of cones  
 originated at  $p_i$

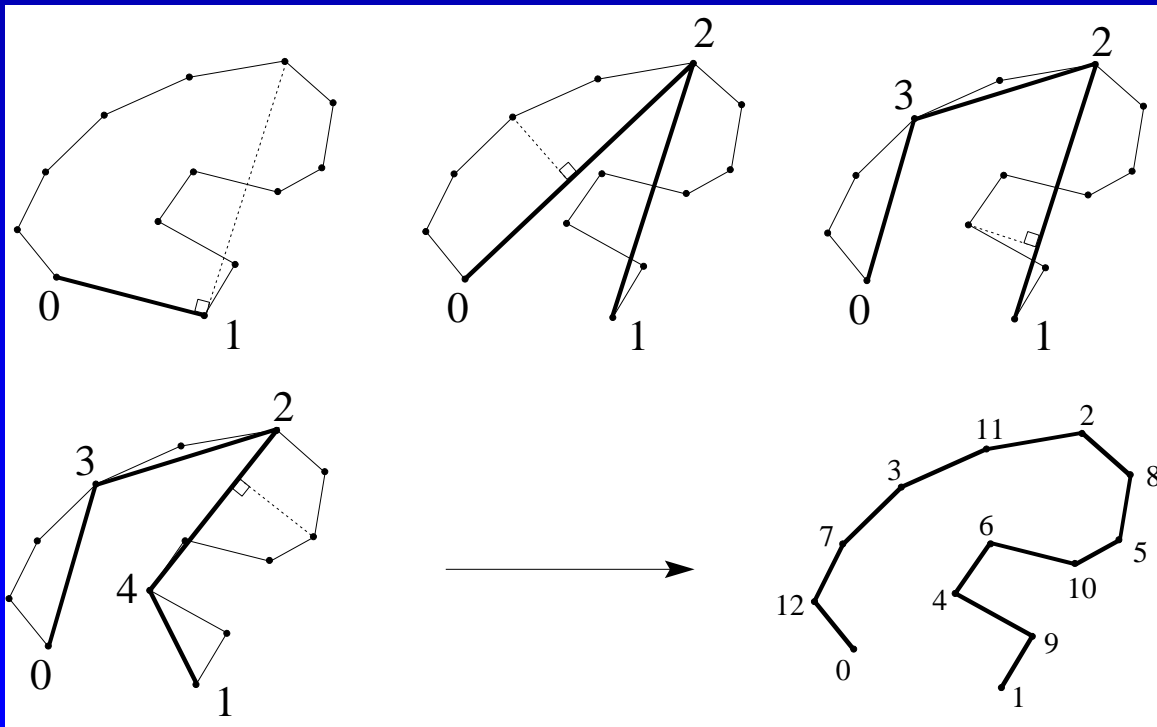
# Survey of the present methods

## ■ Douglas and Peucker's method without tolerance

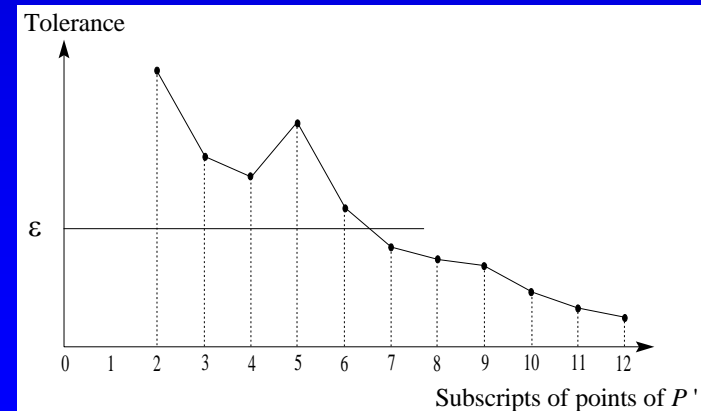
General strategy

Rearrange the points of  $P$

$$P = (p_0, \dots, p_n) \longrightarrow P' = (p'_0, \dots, p'_n)$$



## Multi-scale analysis



# Survey of the present methods

## ↪ Strategies based on spline curves

### ■ The knot removal strategy of T. Lyche & K. Morken

Inputs

⇒ A polygonal curve  $P = (p_0, \dots, p_n)$

⇒ A tolerance  $\varepsilon \geq 0$

⇒ An interpolating B-spline  $f$  on  $T = (t_i)_{i=0}^{n+k}$

Output

⇒ An approximating B-spline  $g$   $\tau = (\tau_i)_{i=0}^{m+k}$   
with  $m < n$  and  $\tau \subset T$  such as  $\|f - g\|_* \leq \varepsilon$

General  
strategy

+ Computation of weights  $\omega_i = \|f - g^i\|_*$

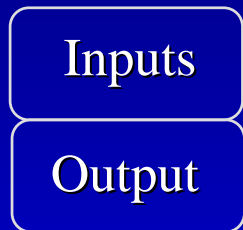
+ Selection of knots to be removed

$t_i$  can be removed if  $\omega_i \leq \varepsilon$

+ Reconstruction of the approximating curve

# Survey of the present methods

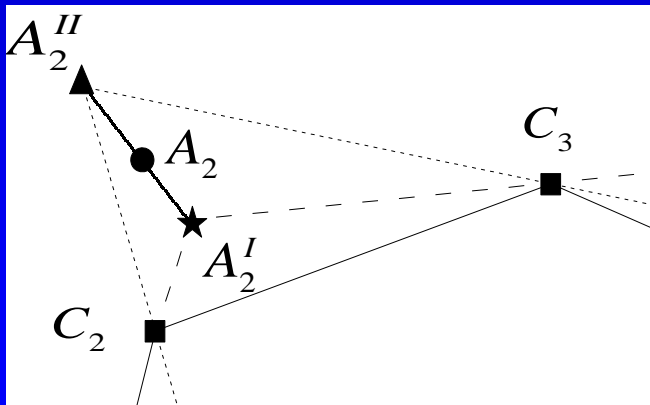
## ■ The knot removal strategy of M. Eck & J. Hadenfeld



The same as in T. Lyche and K. Mørken's strategy  
 where  $f(t) = \sum_{i=0}^n C_i N_{i,k,T}(t)$  is an interpolating B-spline



Removal of knot  $t_i \Rightarrow$  Approximating Bspline  $g^i$  on  $\mu = T - \{t_i\}$



$$g^i(t) = \sum_{j=0}^{n-1} A_j N_{j,k,\mu}(t)$$

"forward" construction  
 of  $g^I$  with  $A_j^I$

"backward" construction  
 of  $g^{II}$  with  $A_j^{II}$

+ Position of control point  $A_j$  in  $[A_j^I, A_j^{II}]$

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- ✉ **A new compression strategy**
  - ↳ Data fitting with B-splines
  - ↳ Criteria for estimating data approximation
  - ↳ Reduction technique
  - ↳ Result comparison
- ✉ Conclusion



# A new compression strategy

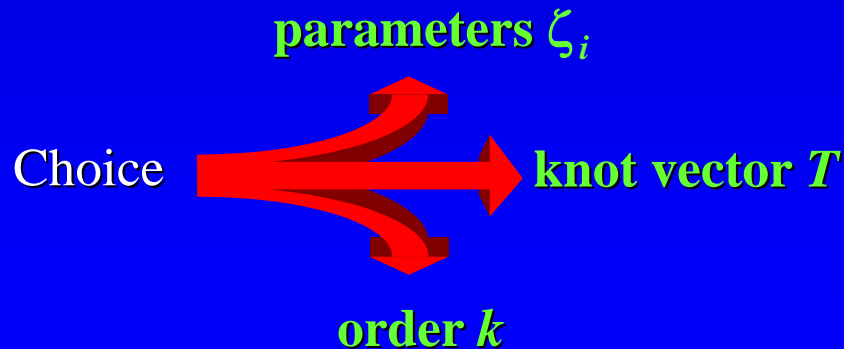
## ↪ Data fitting with B-splines

Problem

Given a polygonal curve  $P = (p_0, \dots, p_n)$ , find a curve  $f(t) = \sum_{i=0}^m Q_i N_{i,k,T}(t)$  as close as possible of  $P$

Least squares fitting: find control points  $Q_j$  so that

$$\sum_{j=0}^n (f(\zeta_j) - p_j)^2 \text{ is minimum} \quad \longrightarrow \quad \text{Householder}$$



# A new compression strategy

## Choice of a parameterization method

→ **uniform**  $h_i = \text{constant}$

→ **cumulative chord length**  $h_i = \|p_{i+1} - p_i\|$

→ **centripetal**  $h_i = \sqrt{\|p_{i+1} - p_i\|}$

## General expression proposed by Lee

$$\zeta_0 = 0, \quad \zeta_i = \frac{\sum_{j=0}^i \|p_j - p_{j+1}\|^e}{\sum_{j=0}^n \|p_j - p_{j+1}\|^e} \quad (e \leq 0 \leq 1)$$

- ▶ uniform with  $e = 0$
- ▶ cumulative chord length with  $e = 1$
- ▶ centripetal with  $e = 0.5$

# A new compression strategy

## Choice of a knot vector

→ **uniform** constant spacing

→ **from an extension of De Boor formula for interpolation**

$$\left\{ \begin{array}{l} t_0 = \dots = t_{k-1} = \zeta_0 \\ t_{m+1} = \dots = t_{m+k} = \zeta_n \\ t_{i+k} = \frac{\zeta_{i+l_1} + \dots + \zeta_{i+l_2}}{l_2 - l_1 + 1} \quad \text{for } i = 0, \dots, m-k \end{array} \right.$$



**Best results in approximation with**

**a knot vector from the extension  
of De Boor formula**

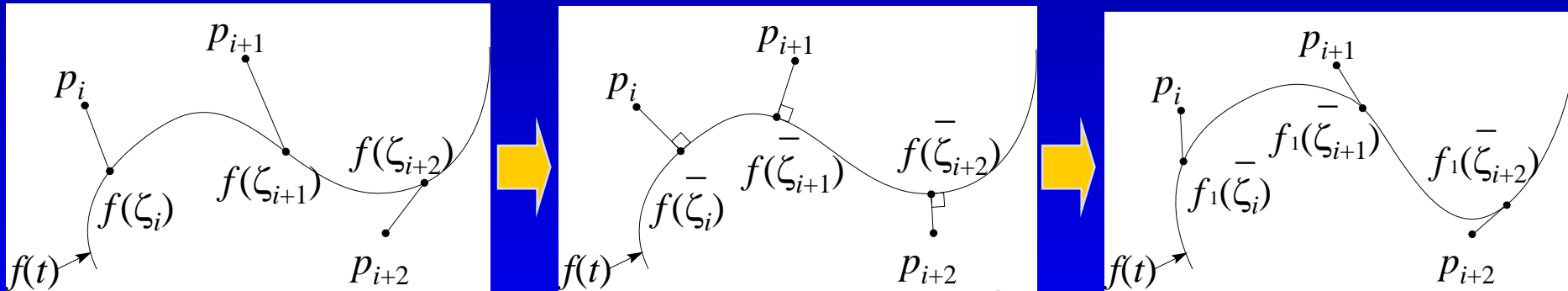
**a centripetal  
parameterization**

# A new compression method

## Two other parameterizations

➔ **Foley and Nielson's parameterization** : geometrical properties (angles)

➔ **Intrinsic Hoschek's parameterization**



### Improvement of Hoschek's method

- Best global approximation
- Best convergence speed

### Common problem

- Oscillation problem

# A new compression method

## Criteria for estimating data approximation

### Definitions of norms : $N_\infty$ et $N_2$

$f(t) = \sum_{i=0}^m Q_i N_{i,k,T}(t)$  is the fitting B-spline curve

- ▶  $k$  is the order
- ▶  $T = (t_0, \dots, t_{k-1}, t_k, \dots, t_m, t_{m+1}, \dots, t_{m+k})$  is the knot vector

$f$  belongs to linear space  $S_{k,T}$

Definitions

$$N_\infty(f) = \text{Max} \left\{ \text{Max} \left\{ |Q_{ji}| ; i = 0, \dots, m \right\} ; j = 1, \dots, d \right\}$$

$$N_2(f) = \frac{\sqrt{\sum_{i=0}^m |Q_i|^2}}{(m+1)}$$

# A new compression method

## Polygonal curve

$g(t) = \sum_{i=0}^n R_i N_{i,2,\tilde{T}}(t)$  is the interpolating B-spline curve of initial data  $p_i$

▶ order = 2

▶  $\tilde{T} = (\tilde{t}_0, \dots, \tilde{t}_{k-1}, \tilde{t}_k, \dots, \tilde{t}_m, \tilde{t}_{m+1}, \dots, \tilde{t}_{m+k})$  is the knot vector

**$g$  belongs to linear space  $S_{2,\tilde{T}}$**

**To use  $N_\infty$  and  $N_2$  on  $f-g$ , B-spline curve  $f$  and  $g$  should have**

▶ **the same order**

degree elevation (H. Prautzsch)

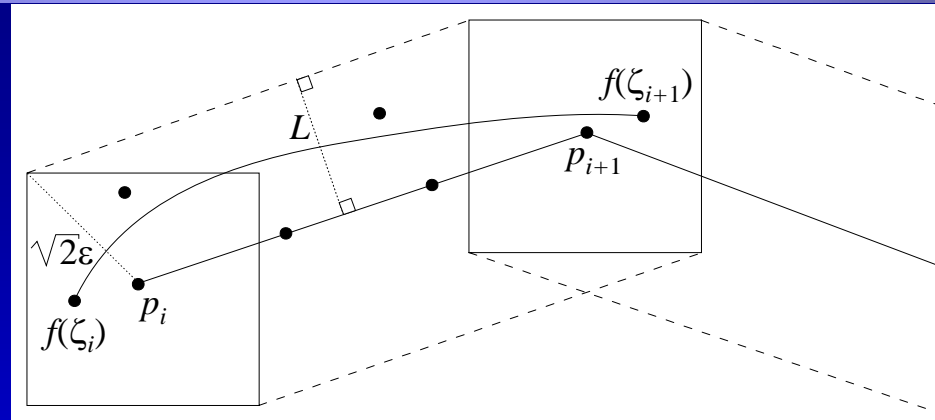
▶ **the same knot vector**

subdivision algorithms (Boehm, Oslo, improved Oslo)

# A new compression strategy

Result

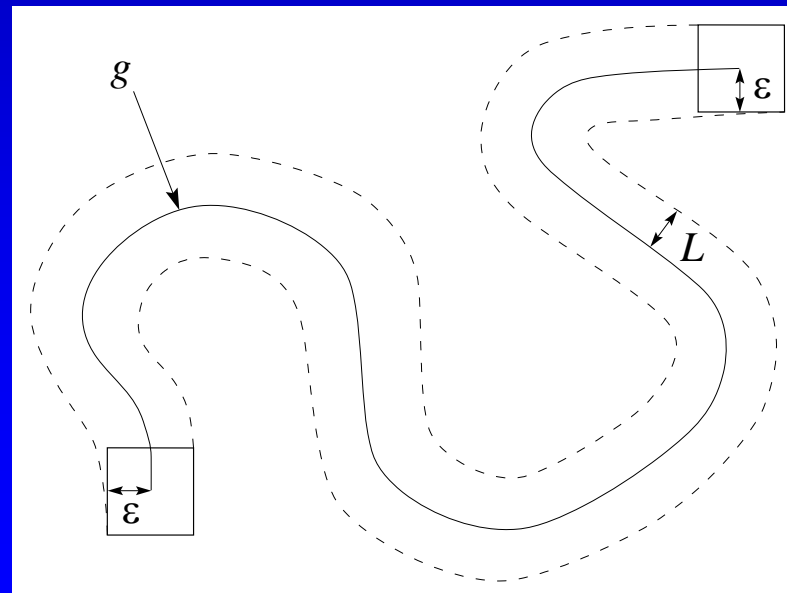
**Band criterion  
for data approximation**



$$N_{\infty}(f - g) = \varepsilon$$

Result

**Band criterion  
for B-spline approximation**



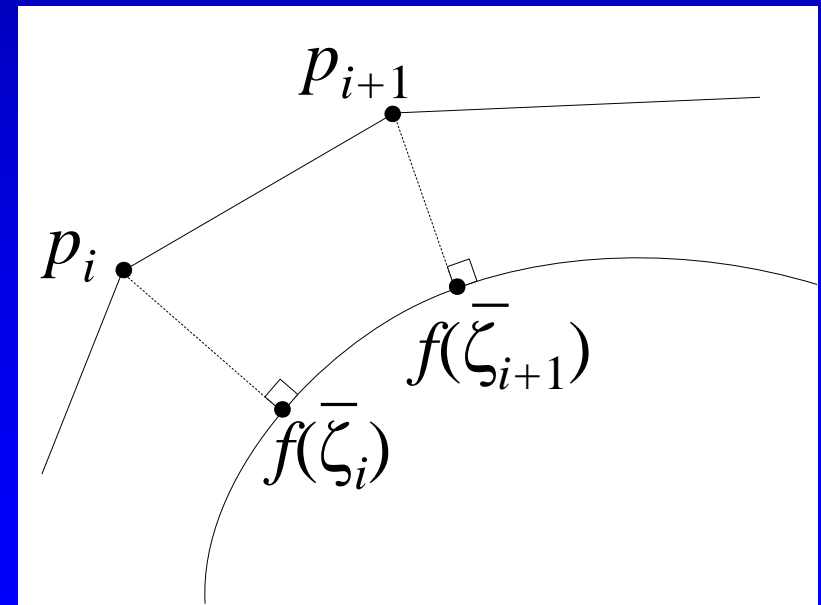
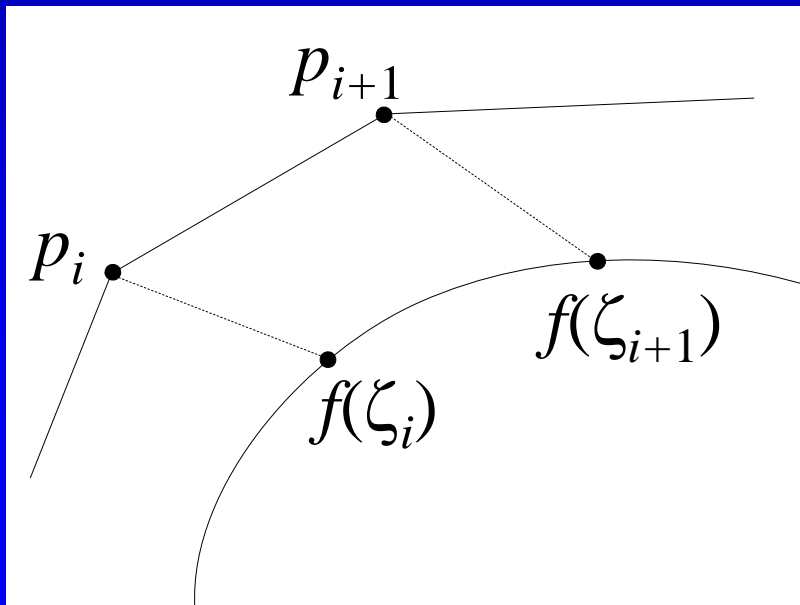
# A new compression strategy

## Local estimation criteria

Aim

To have a **local approach of approximation and estimation**

+ Find the best curve segment  $[p_i, p_{i+1}]$

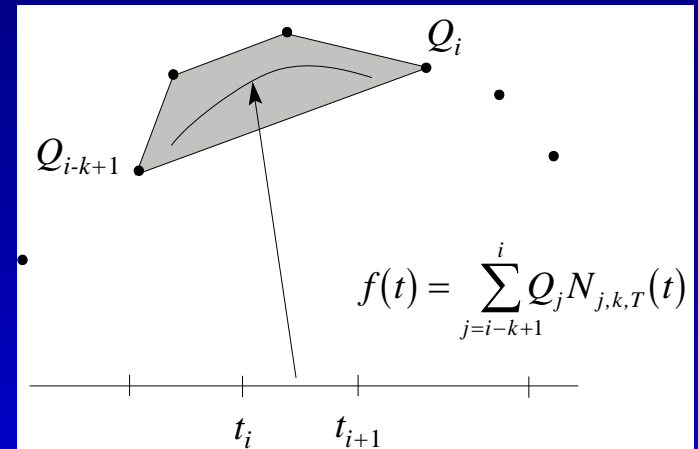




# A new compression strategy

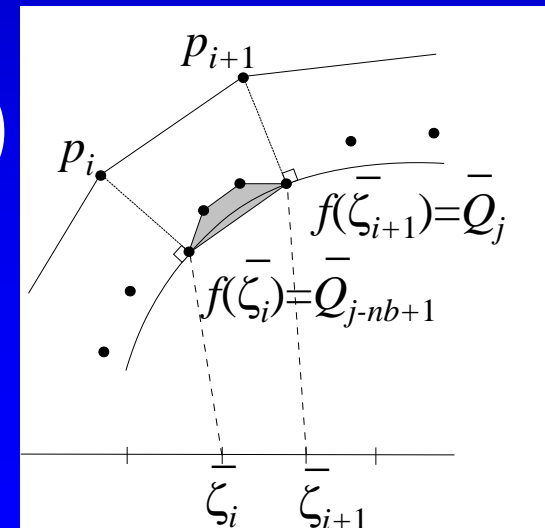
+ Use the convex hull of B-splines

NOT ENOUGH



+ Transform this best curve segment into its "Bézier" representation

Apply the Hausdorff metric on both polygonal curves  $P^i = (p_i, p_{i+1})$  and  $\bar{Q}^j = (\bar{Q}_{j-nb+1}, \dots, \bar{Q}_j)$



Result

If  $d_H = (P^i, \bar{Q}^j) = \varepsilon$  then the best curve segment is at the most at  $\varepsilon$  distance from  $[p_i, p_{i+1}]$

# A new compression strategy

## ↪ Reduction technique



### ■ Best choices

- ↪ **Uniform knot vector**
- ↪ **Intrinsic Hoschek's parameterization applied to a centripetal one**

### ■ Principle

- ↪ **A bisection method on the number of control points**

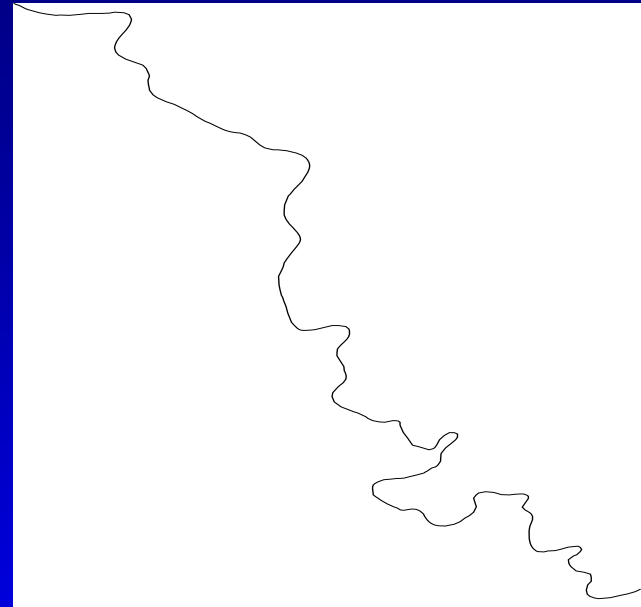
# A new compression strategy

## ↪ Results comparison

### ▣ on compression rate

Initial isobathymetric line

Tolerance so that no visual difference

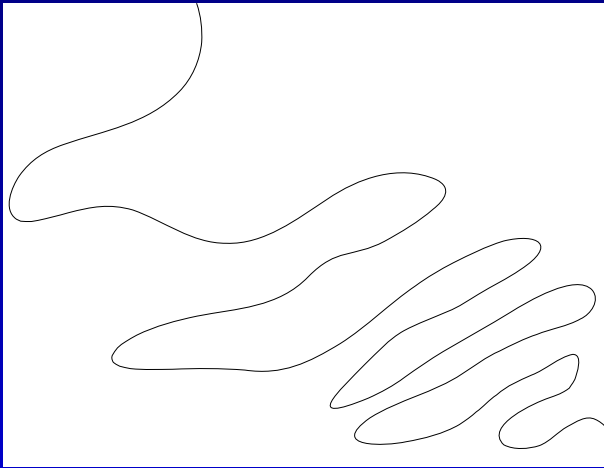


Results

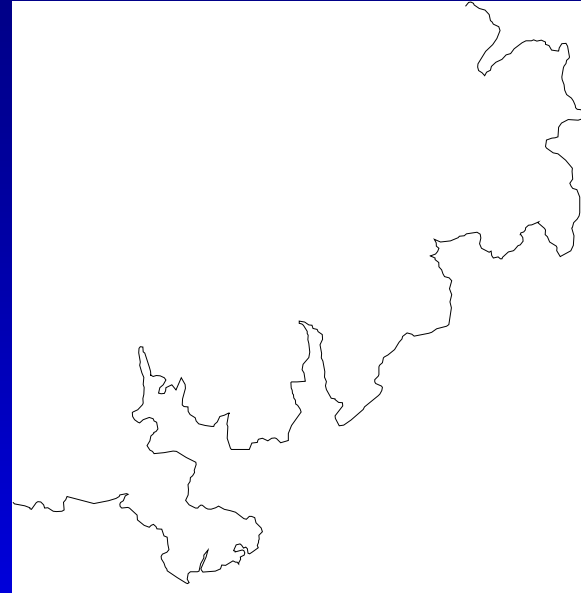
Method	Knots	Strategy	Compression
Polygonal		Arge Daehlen	<b>68 %</b>
		Douglas Peucker	56 %
B-spline	De Boor	Lyche Morken	13 %
		Eck Hadenfeld	51 %
		Fitting ( $e=0.5$ )	22 %
	Uniform	Fitting (Hoschek $e=0.5$ )	<b>61 %</b>
		Fitting ( $e=0.5$ )	49 %

# A new compression strategy

Initial  
smooth line



Initial  
complex line



Results

Method	Knots	Strategy	Compression (Smooth curves)	Compression (Complex curves)
Polygonal		Arge Daehlen	<b>74 %</b>	<b>61 %</b>
		Douglas Peucker	65 %	<b>51 %</b>
B-spline	De Boor	Lyche Morken	24 %	5 %
		Eck Hadenfeld	58 %	38 %
		Fitting ( $e=0.5$ )	34 %	11 %
	Uniform	Fitting (Hoschek $e=0.5$ )	<b>68 %</b>	<b>49 %</b>
		Fitting ( $e=0.5$ )	59 %	40 %

# A new compression strategy

## ↪ Results comparison

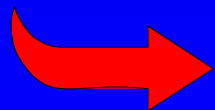
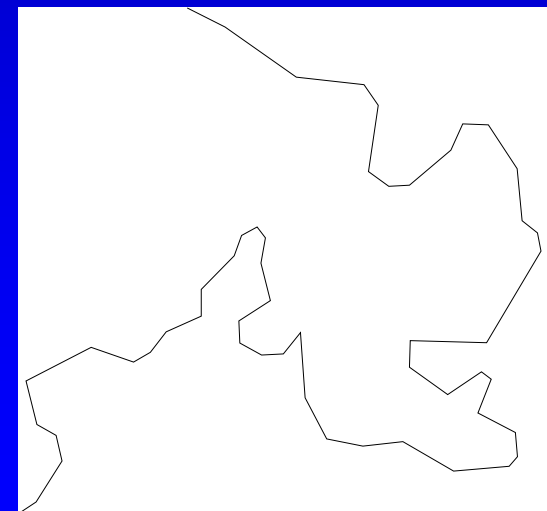
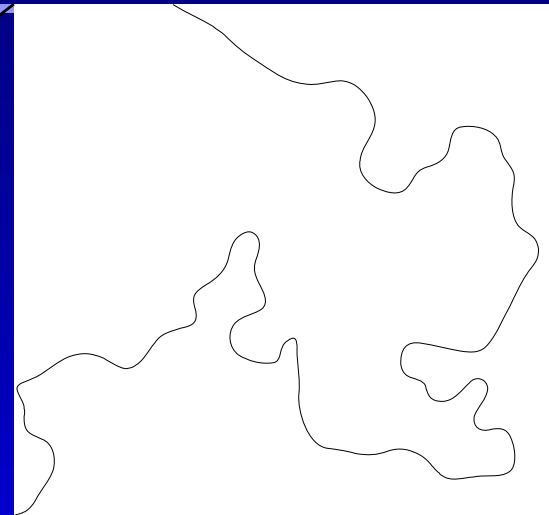
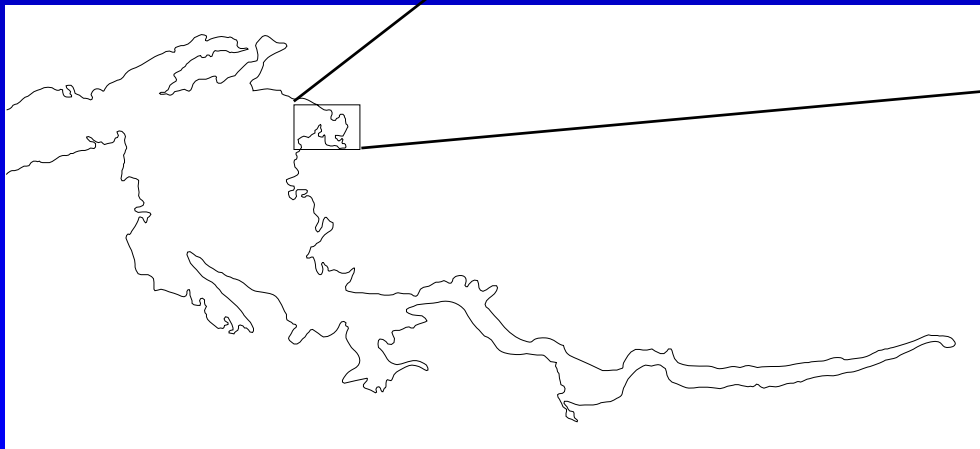
### ▣ on computation time

#### Results

Method	Knots	Strategy	Computation time (Smooth curves)	Computation time (Complex curves)
Polygonal		Arge Daehlen	<b>0.1</b> ''	<b>2</b> ''
		Douglas Peucker	<b>0.4</b> ''	<b>3</b> ''
B-spline	De Boor	Lyche Morken	47' 08''	1h 22''
		Eck Hadenfeld	40' 31''	59' 21''
		Fitting ( $e=0.5$ )	1' 27''	3' 27''
	Uniform	Fitting (Hoschek $e=0.5$ )	9' 33''	13' 24''
		Fitting ( $e=0.5$ )	1' 52''	4' 45''

# A new compression strategy

↪ **Advantage when**  
▣ **zooming in**



**Useful in embarked**  
**cartographic information systems**

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# Conclusion

We have **build**

➔ **Norms  $N_\infty$  and  $N_2$**

➔ **Local estimation criteria**

We have **improved**

➔ **Hoschek's technique**

Which has allowed us to **introduce**

➔ **A new reduction strategy**

- ▶ Good reduction rates
- ▶ Reasonable computation costs



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