



MLOMA: Machine Learning, Optimization and Manifolds

21-21 Dec 2023 Clermont-Ferrand (France)

Two “non-standard” methods to analyze 2D/3D shapes

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2D SHAPE ANALYSIS BASED ON PLANAR MECHANISM DESIGN

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Thanks to José Braga (CAGT, Toulouse, France) and Guillaume Captier (ICAR, LIRMM) for providing data

In general, based on Fourier descriptors:

- Is it well adapted to the shape?
20 harmonics gives $4 \times 20 = 80$ coefficients to explain the positions of 84 points
- EEF parameters are not directly related to local geometry
- May be sensitive to the definition of the origin point.

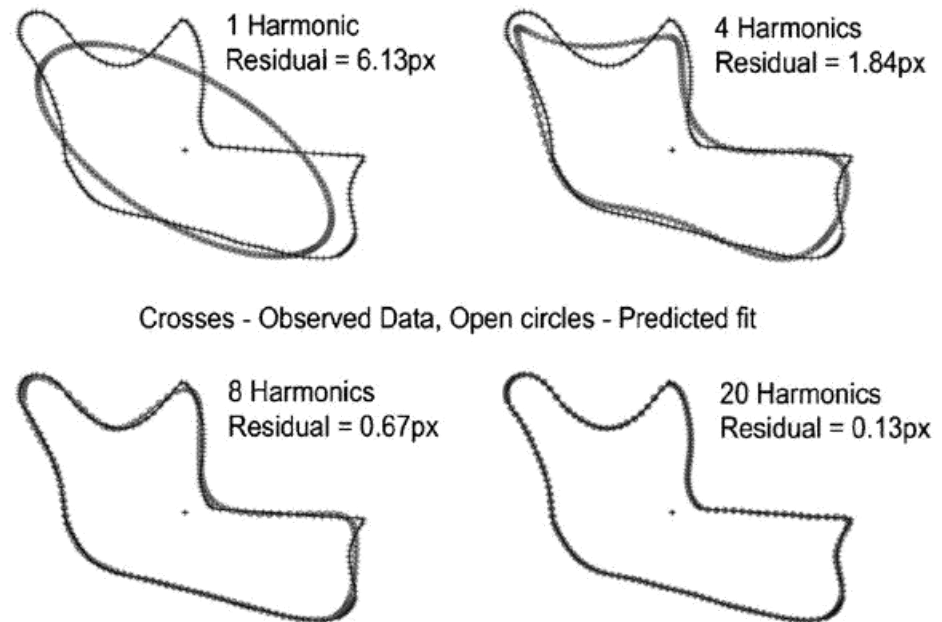


Fig. 6. Convergence of the EEF fit to the mandible outline.

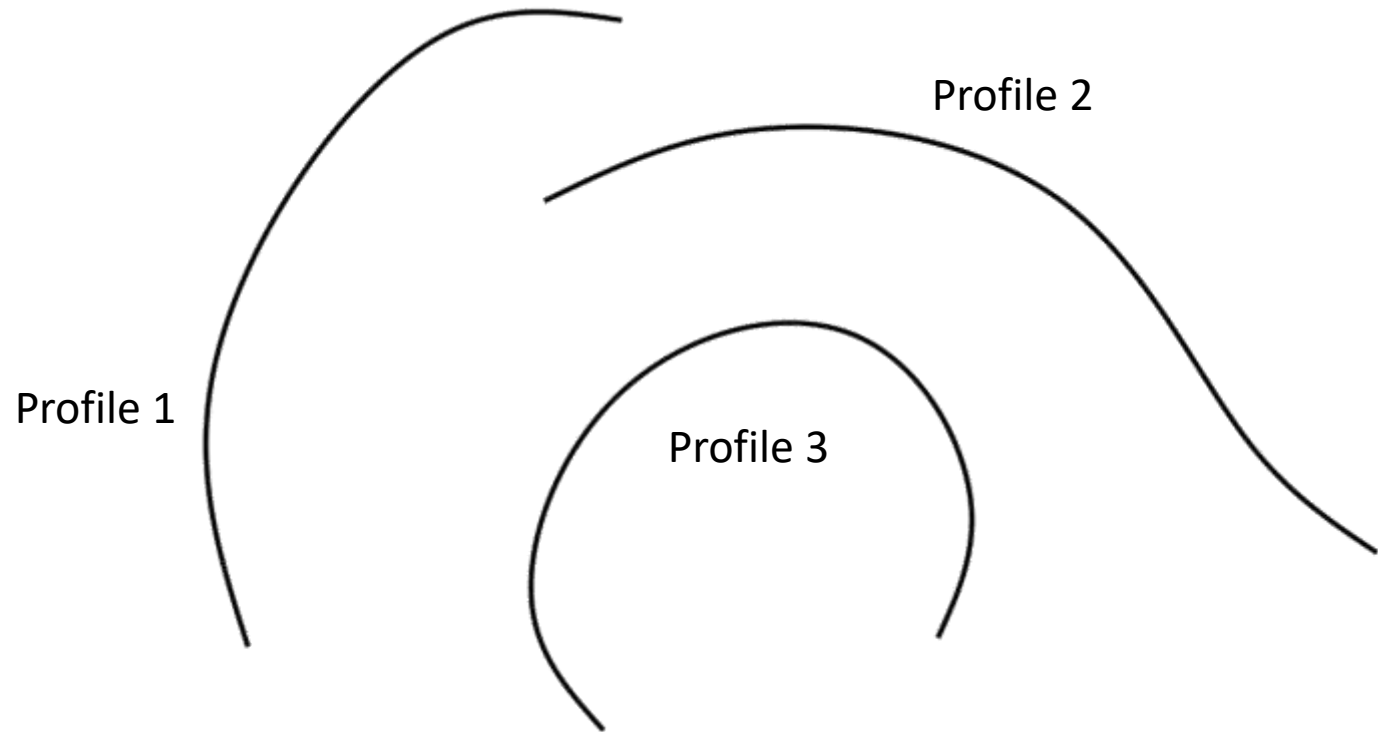
- New method based on mechanical considerations.....
- Shape changing rigid-body mechanisms
- Mechanism: revolute joints (pivot) / prismatic joints (glissière) which parameters are easy to understand.

*B. Li, A.P. Murray, D.H. Myszka, G. Subsol.
"Synthesizing Planar Rigid-Body Chains for
Morphometric Applications". ASME
International Design Engineering Technical
Conferences & Computers and Information in
Engineering Conference, Charlotte (U.S.A.),
August 2016.*



The Goal

Different profile shapes



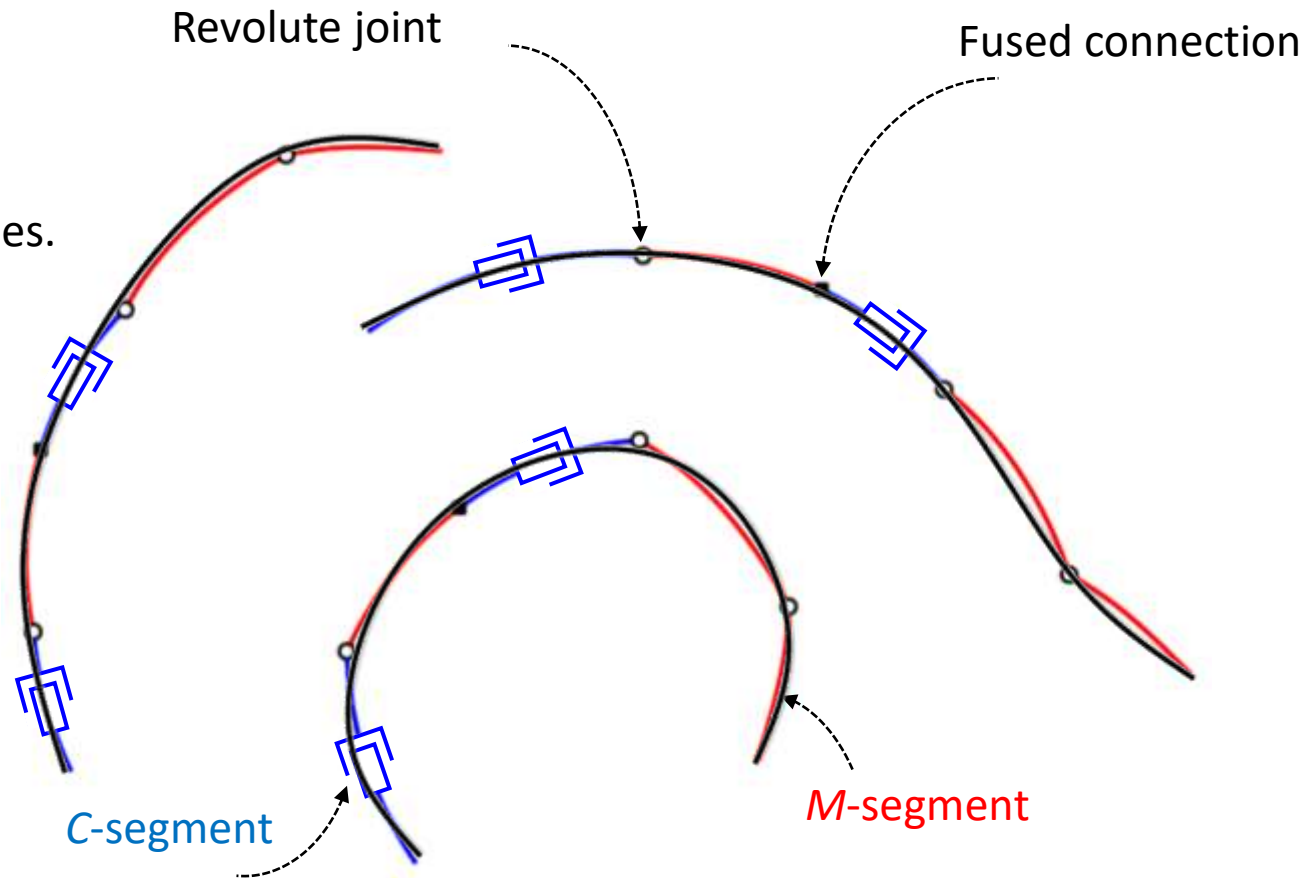
Find a chain of rigid bodies composed of:

- Constant-curvature segments (slide)
- **Mean fixed segments**

connected by:

- Fused connections
- Revolute joints

which fits with all the profiles.

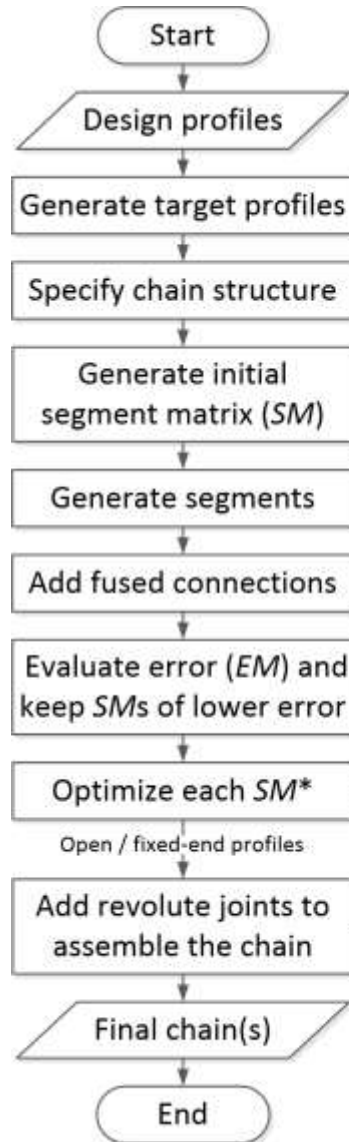


C/R/M/F/C/R/M/R/M

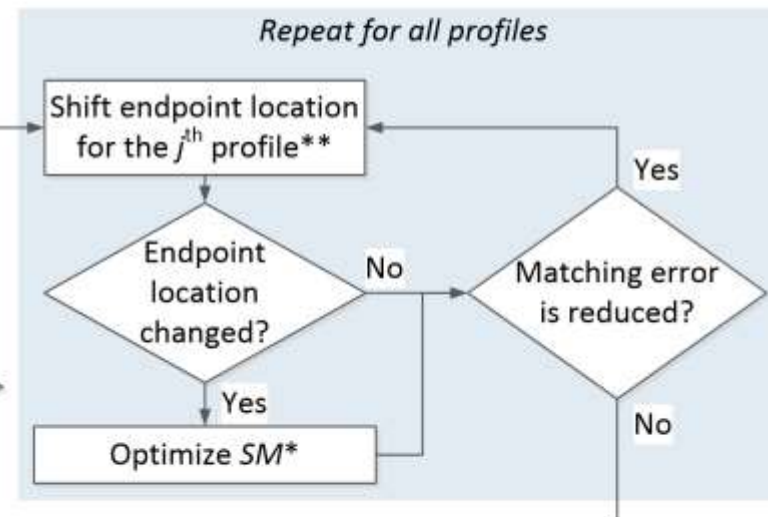
The Goal

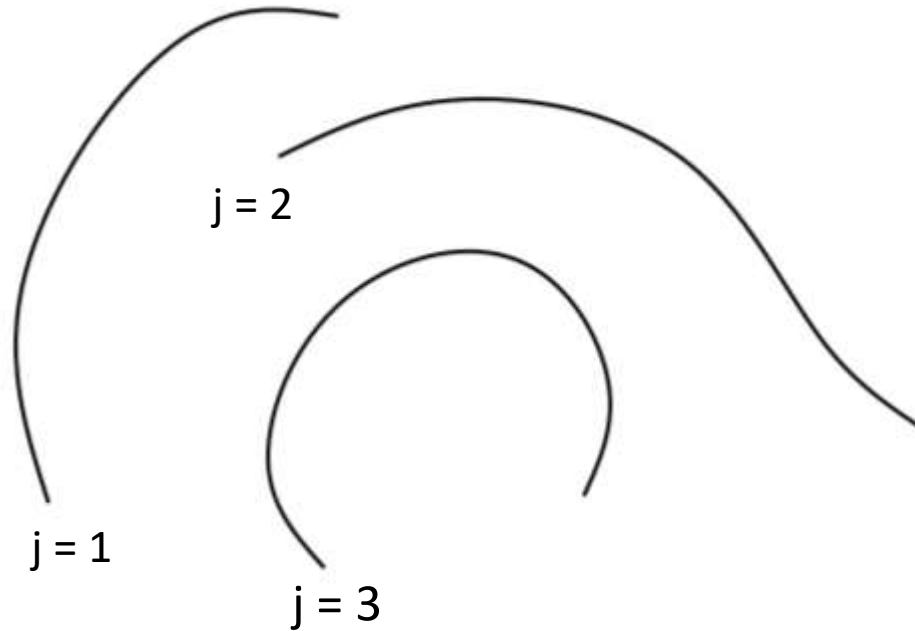
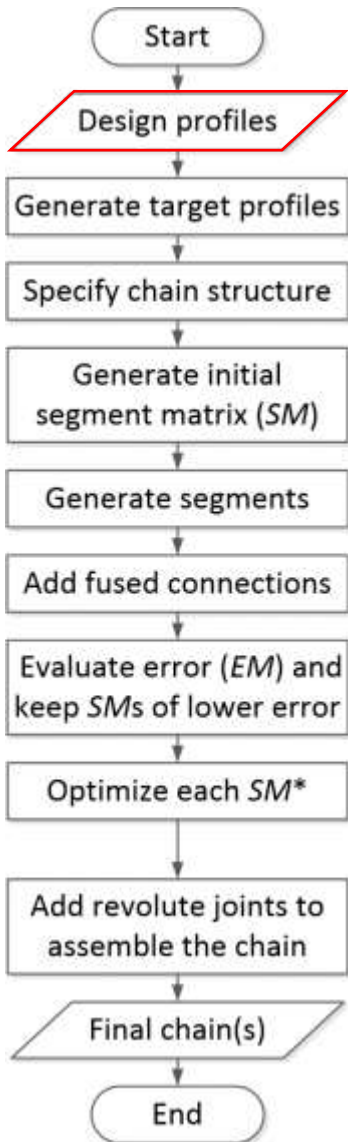


2 **C segments** + 3 **R joints** = 5 scalar parameters (angles) only by profile!

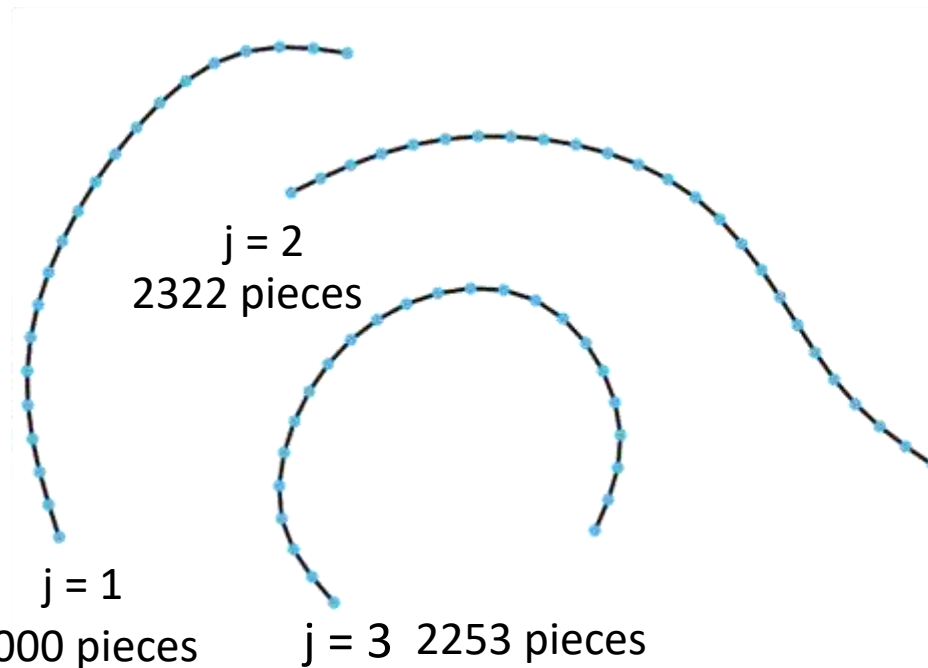
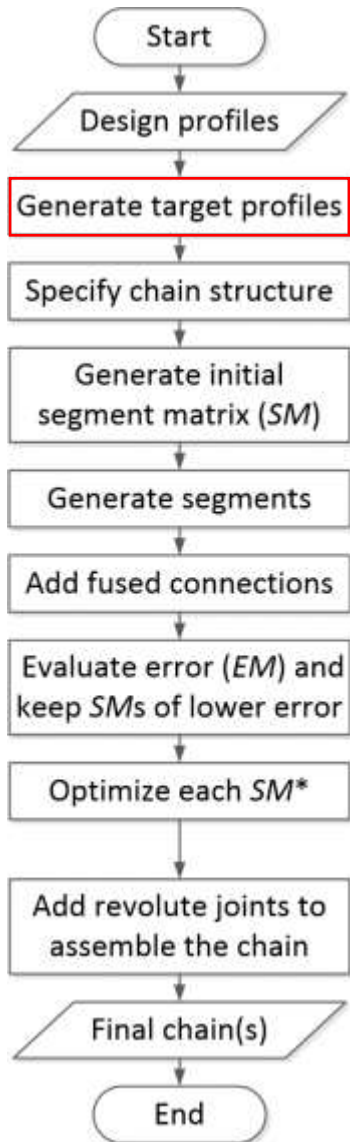


Li, B., Murray, A., Myszka, D., and Subsol, G., "Synthesizing planar rigid-body chains for morphometric applications", submitted to the 2016 ASME International Design and Engineering Technical Conferences, Charlotte, NC, Aug. 21-24, 2016





- Can be generated from:
- Mathematical functions
 - Point coordinates
 - Line drawings



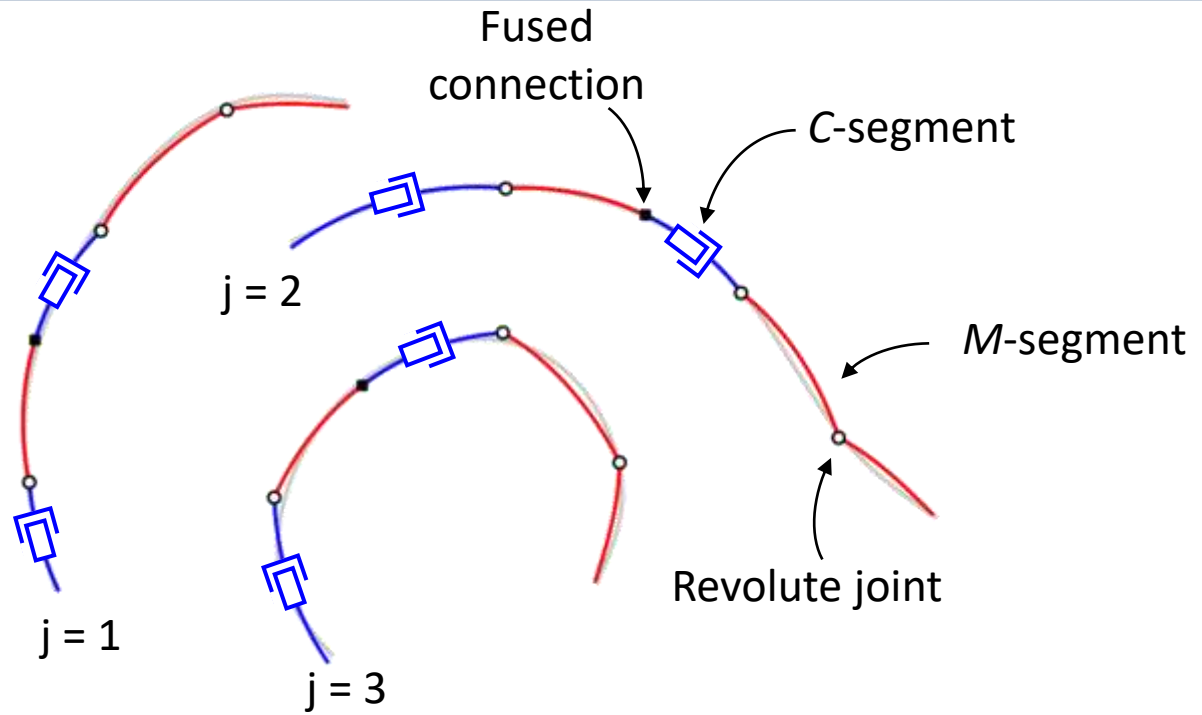
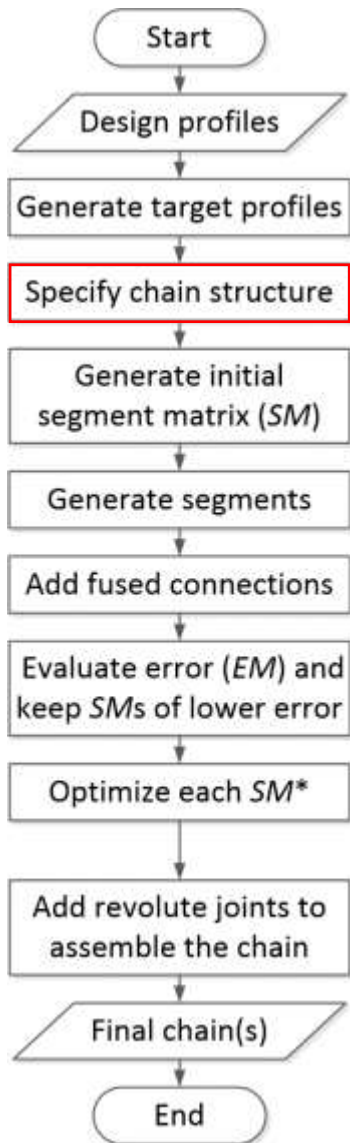
1. Compute the arc length of each design profile
2. Determine the number of pieces should be in each target profile
3. Desired piece length of target profile

$$C_j = \sum_{i=1}^{N_j-1} c_{ji}$$

$$m_j = \left\lceil \frac{C_j}{C_{\min}} m_d \right\rceil$$

$$s_j = \frac{C_j}{m_j}$$

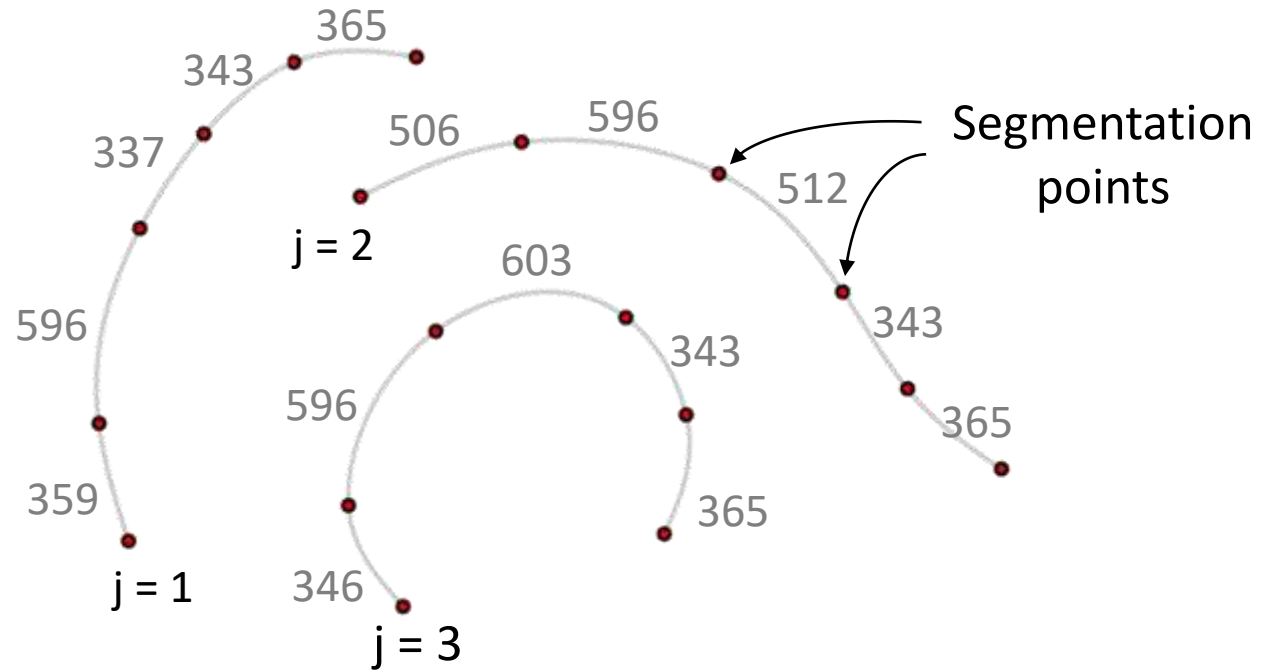
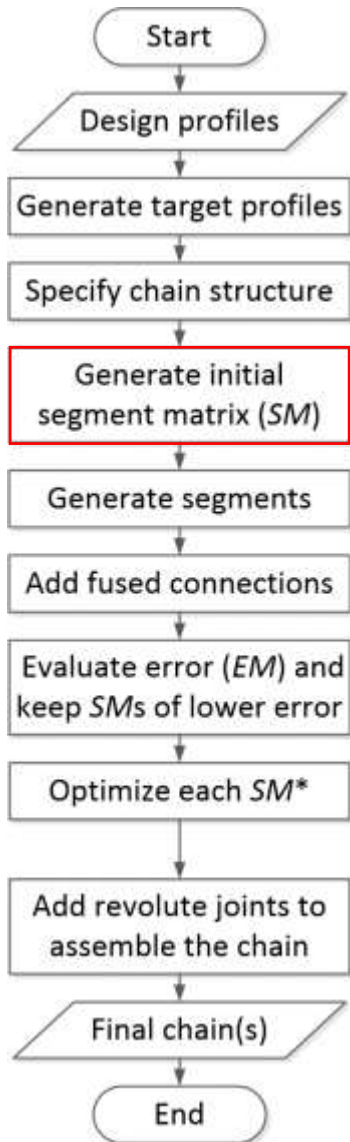
Specify Chain Structure



Segment vector, $\mathbf{V} = [C \ M \ C \ M \ M]$

Connection vector, $\mathbf{W} = [R \ F \ R \ R]$

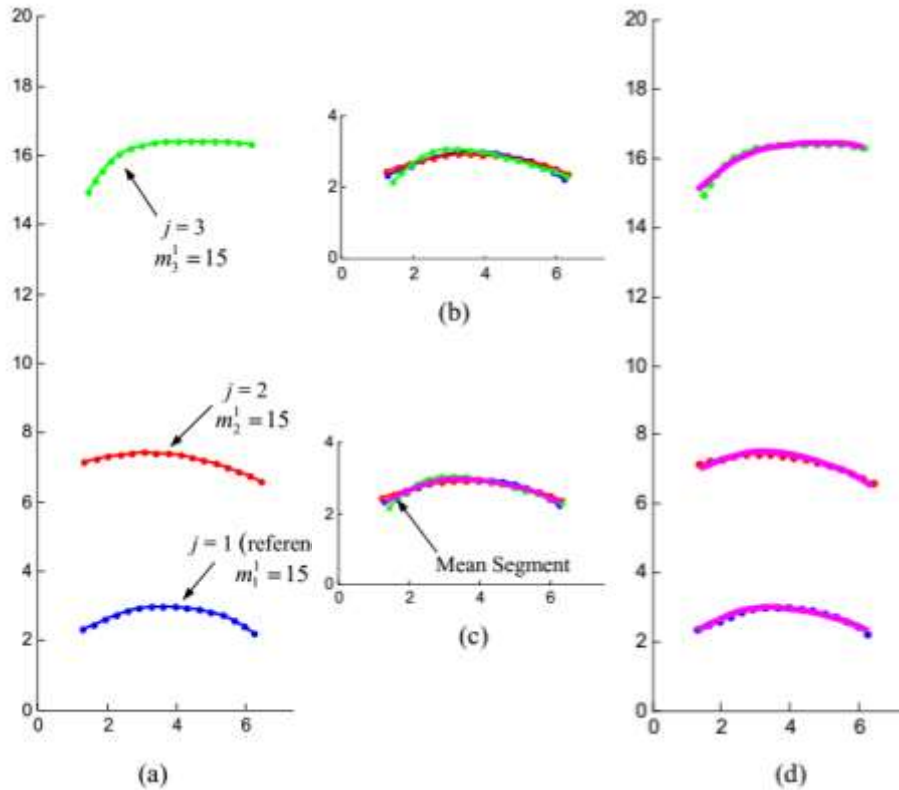
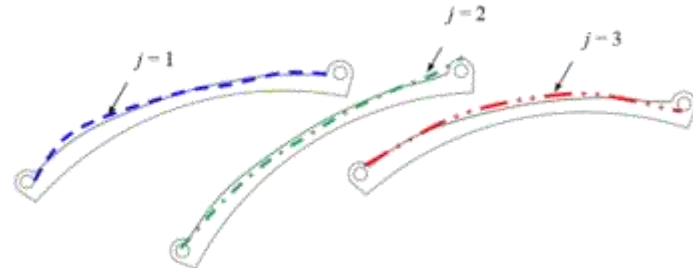
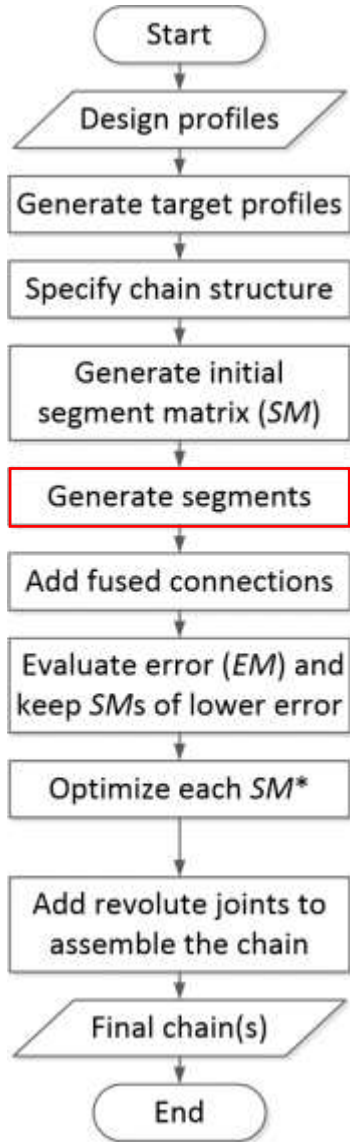
Minimum number of pieces in a segment, α



$$\mathbf{V} = [C \quad M \quad C \quad M \quad M]$$

$$SM = \begin{bmatrix} 359 & 596 & 337 & 343 & 365 \\ 506 & 596 & 512 & 343 & 365 \\ 346 & 596 & 603 & 343 & 365 \end{bmatrix} \begin{matrix} j = 1 \\ j = 2 \\ j = 3 \end{matrix}$$

M-Segments



Finding average geometry

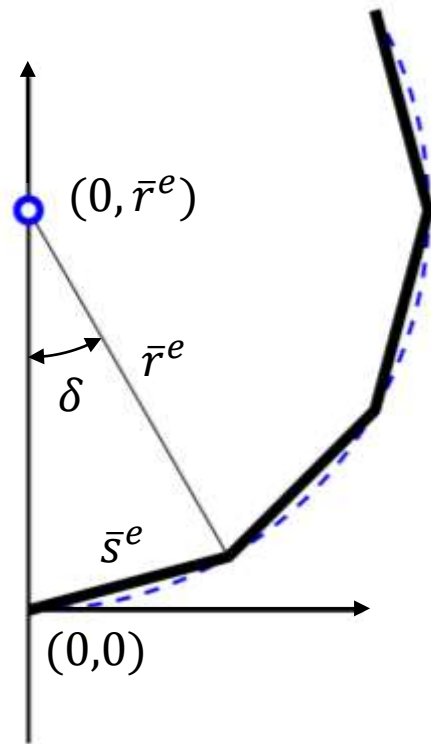
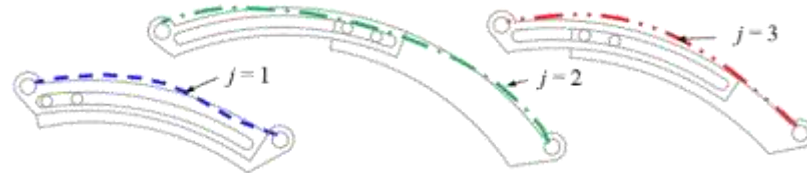
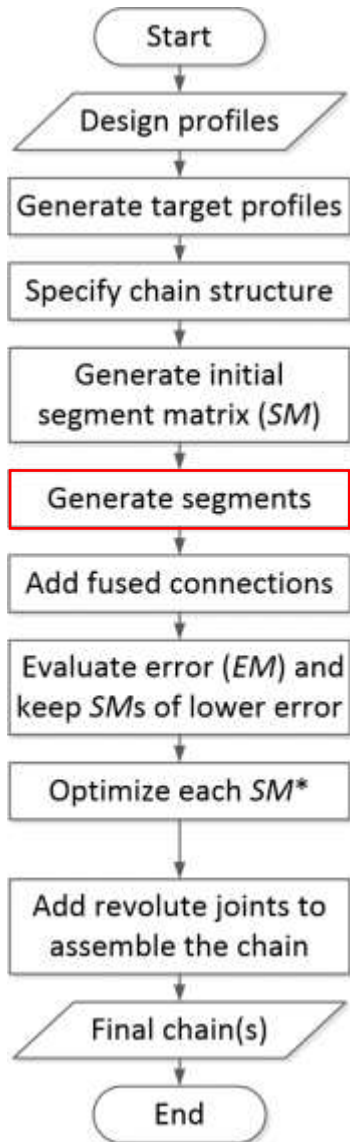
$$\mathbf{Z}_{m_i} = \frac{1}{p} \left(\mathbf{z}_{1_i} + \sum_{j=2}^p \mathbf{Z}_{j_i} \right)$$

Actual segment point coordinates

$$\bar{\mathbf{z}}_{j_i}^e = \left(\hat{\mathbf{A}}_j^e \right)^{-1} \left(\mathbf{Z}_{m_i} - \hat{\mathbf{d}}_j^e \right)$$

$$i = k_j^e, \dots, k_j^{e+1} \quad j = 2, \dots, p.$$

C-Segments



Average radius (reciprocal of curvature)

$$\bar{r}^e = \frac{1}{\sum_{j=1}^p m_j^e - 1} \left(\sum_{j=1}^p \left(\sum_{i=k_j^e+1}^{k_j^e+1} r_{ji} \right) \right)$$

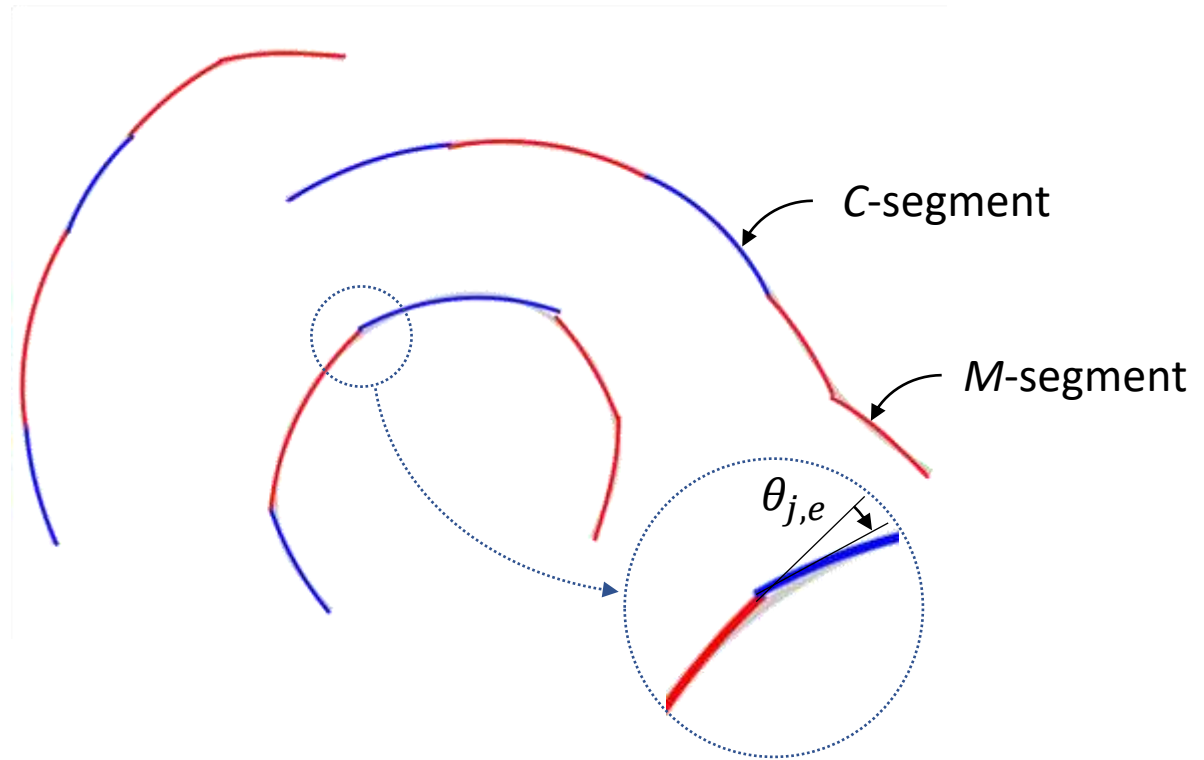
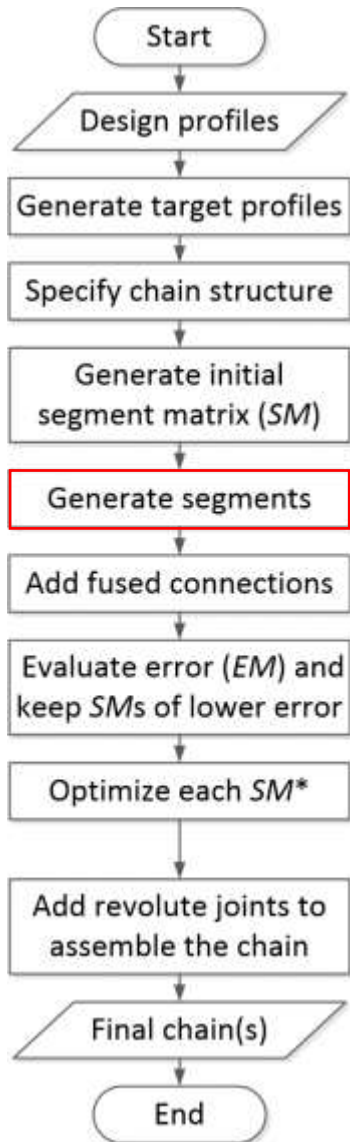
Segment arc length

$$L_j^e = m_j^e \bar{s}^e$$

Average piece length

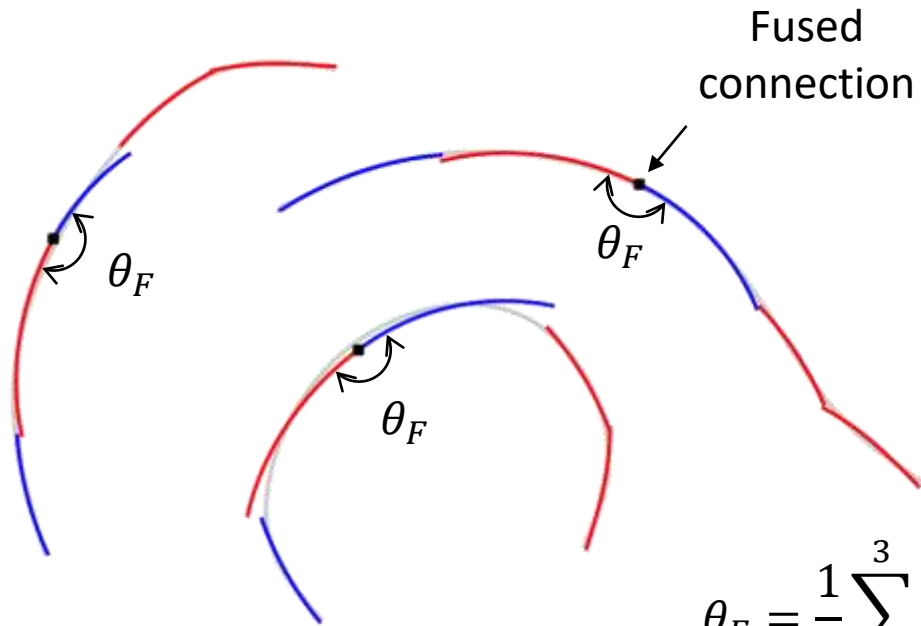
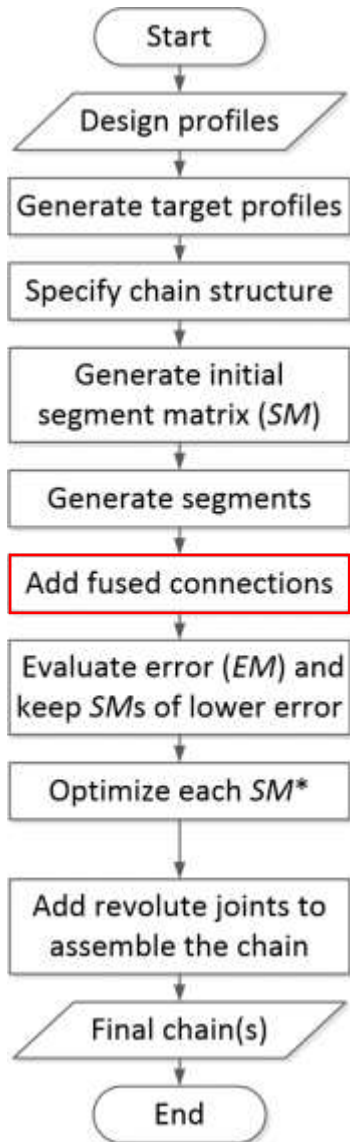
$$\bar{s}^e = \sum_{j=1}^p \left(\frac{1}{m_j^e} \left(\sum_{i=k_j^e+1}^{k_j^e+1} s_{ji} \right) \right)$$

Generate Segments



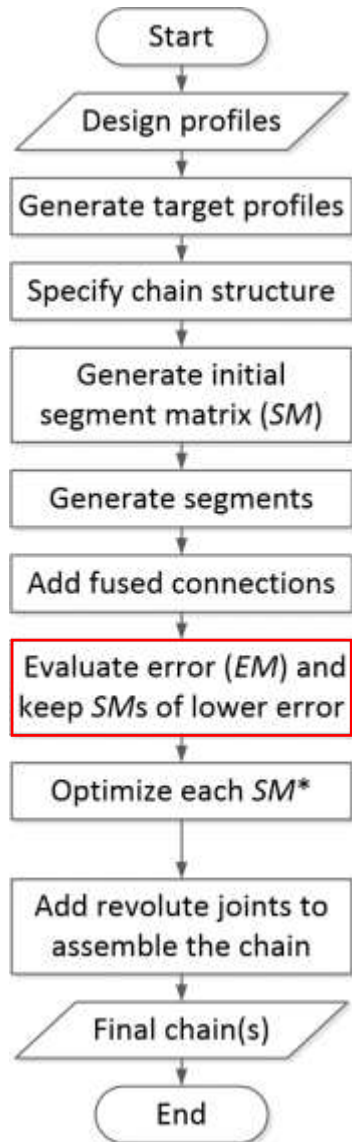
$$\mathbf{V} = [C \quad M \quad C \quad M \quad M]$$

$$SM = \begin{bmatrix} 359 & 596 & 337 & 343 & 365 \\ 506 & 596 & 512 & 343 & 365 \\ 346 & 596 & 603 & 343 & 365 \end{bmatrix} \begin{matrix} j = 1 \\ j = 2 \\ j = 3 \end{matrix}$$



$$\theta_F = \frac{1}{3} \sum_{j=1}^3 \theta_{j,2}$$

$$W = [R \mathbf{F} R R]$$



$$E_j^e = \max |\bar{z}_{j_i}^e - z_{j_i}| \quad i = k_j^e, \dots, k_j^{e+1}$$

$$\bar{E} = \frac{1}{q} \sum_{e=1}^q \left(\frac{1}{p} \sum_{j=1}^p E_j^e \right)$$

$$\bar{E}^e = \sum_{j=1}^p E_j^e$$

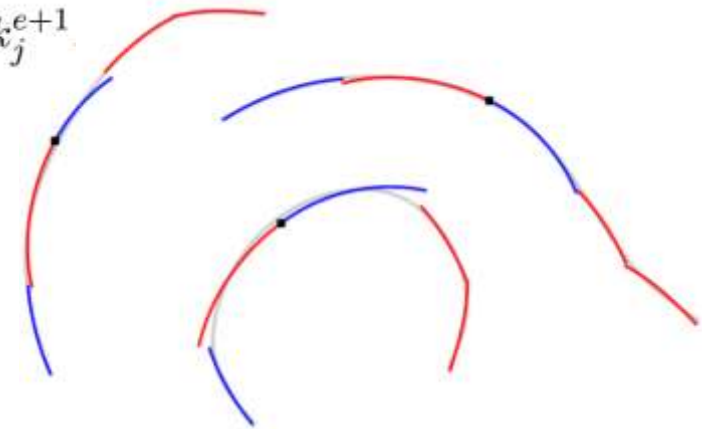
	C	M	C	M	M
E^1	0.35	1.00	1.96	0.27	0.52
E^2	0.46	1.19	1.19	0.78	1.27
E^3	0.81	3.23	4.31	0.55	0.76

Overall mean error,

$$\bar{E} = 1.24$$

Mean error of each M-segment:

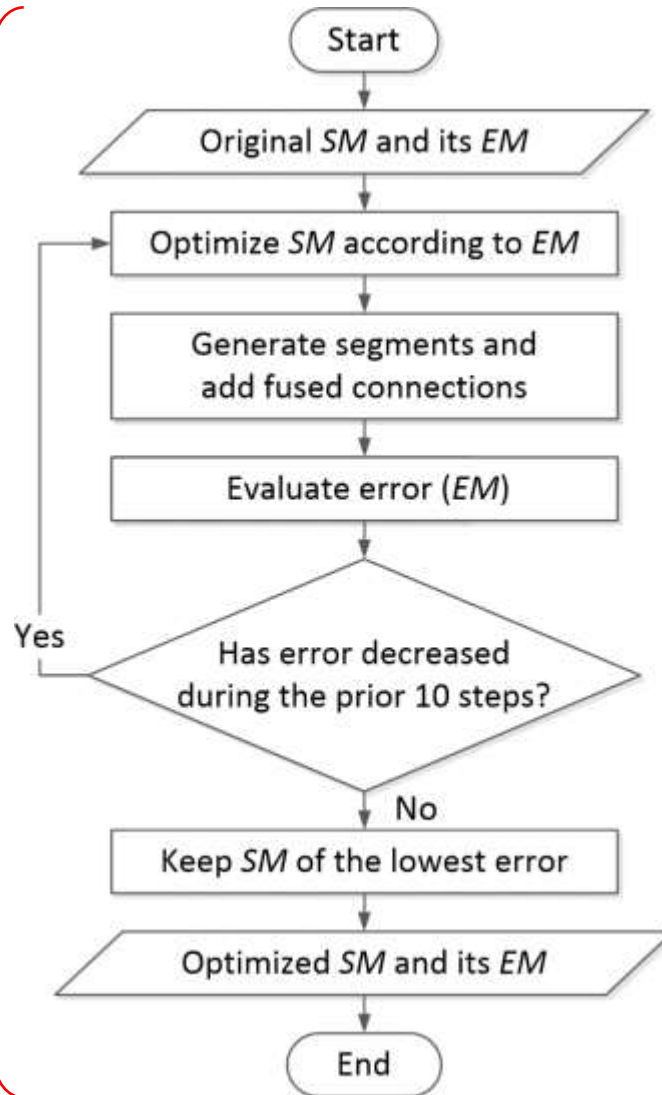
$$\bar{E}^2 = 1.81, \quad \bar{E}^4 = 0.53, \quad \bar{E}^5 = 0.86$$



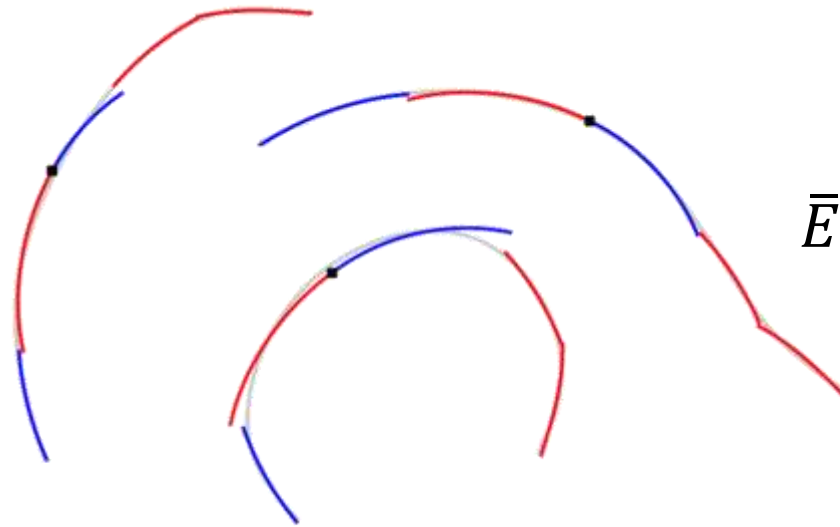
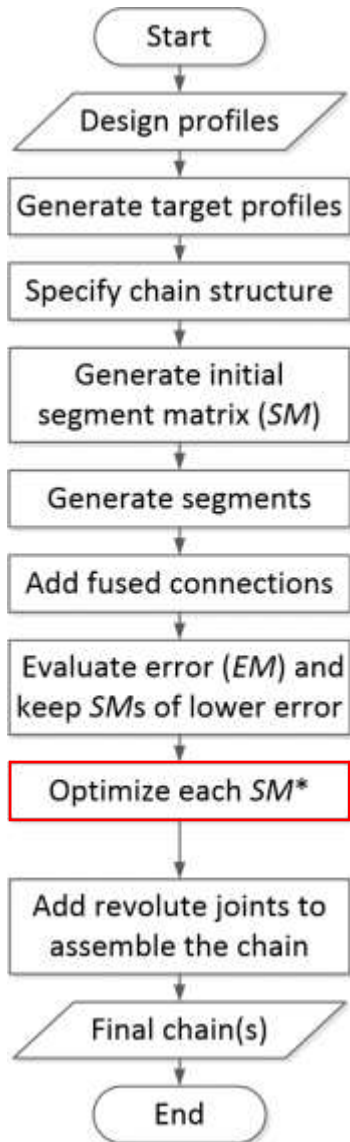
Segment to be elongated
 Segment to be shortened

SM Optimization

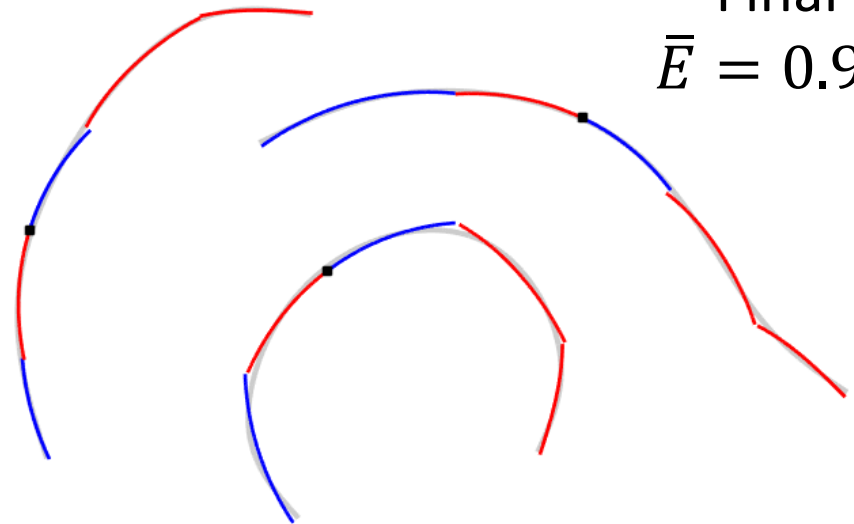
$$\begin{array}{c}
 C \quad M \quad C \quad M \quad M \\
 SM_0 = \begin{bmatrix} 359 & 596 & 337 & 343 & 365 \\ 506 & 596 & 512 & 343 & 365 \\ 346 & 596 & 603 & 343 & 365 \end{bmatrix} \\
 \\
 SM_1 = \begin{bmatrix} 360 & 595 & 335 & 344 & 366 \\ 507 & 595 & 510 & 344 & 366 \\ 346 & 595 & 602 & 344 & 366 \end{bmatrix} \\
 \\
 \vdots \\
 \\
 SM_f = \begin{bmatrix} 326 & 420 & 376 & 522 & 356 \\ 660 & 420 & 364 & 522 & 356 \\ 509 & 420 & 446 & 522 & 356 \end{bmatrix}
 \end{array}$$



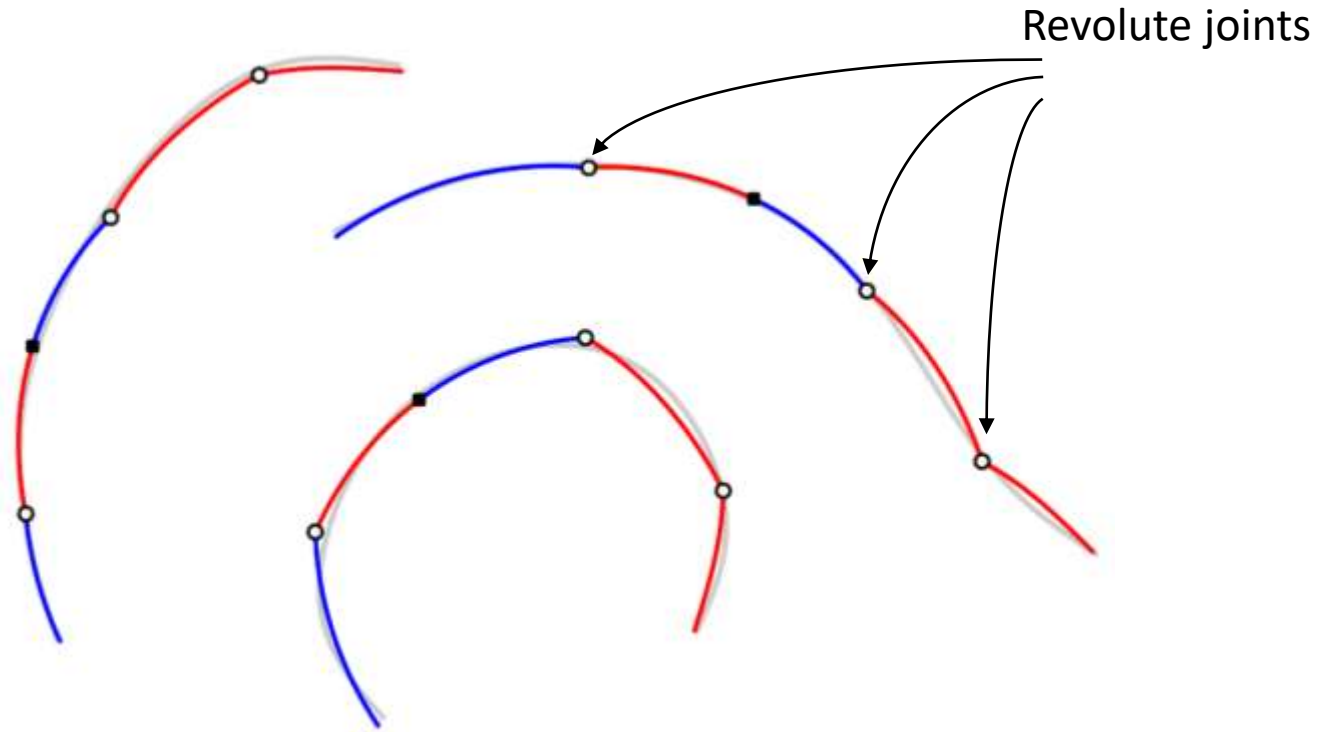
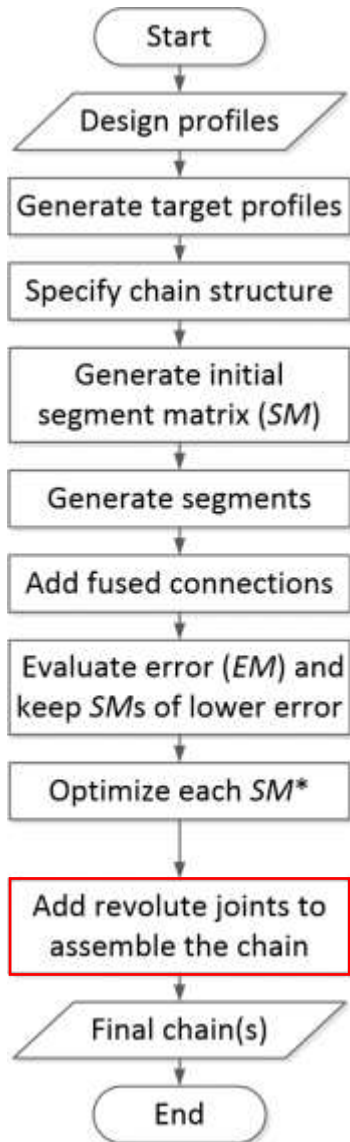
$$\begin{array}{c}
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 \\
 \vdots \\
 \\
 SM_f = \begin{bmatrix} 326 & 420 & 376 & 522 & 356 \\ 660 & 420 & 364 & 522 & 356 \\ 509 & 420 & 446 & 522 & 356 \end{bmatrix}
 \end{array}$$



Initial
 $\bar{E} = 1.24$



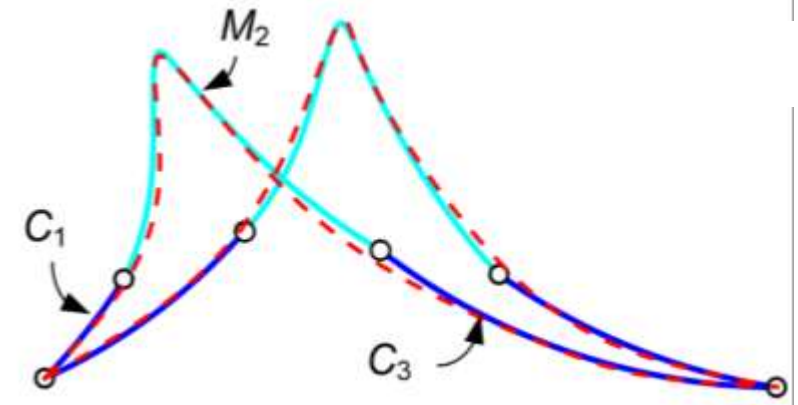
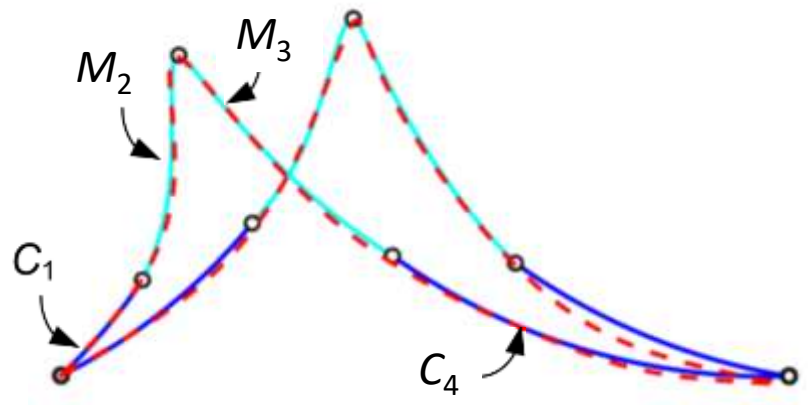
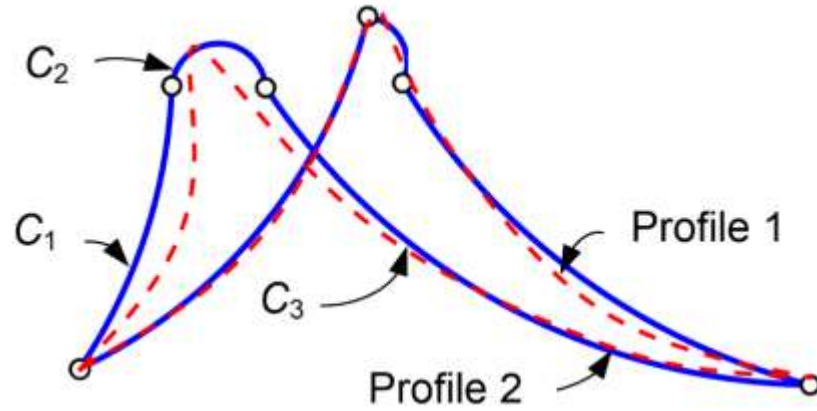
Final
 $\bar{E} = 0.98$

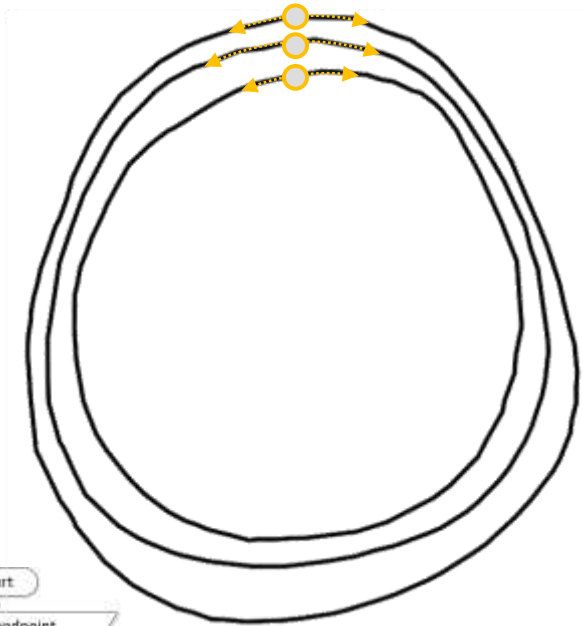
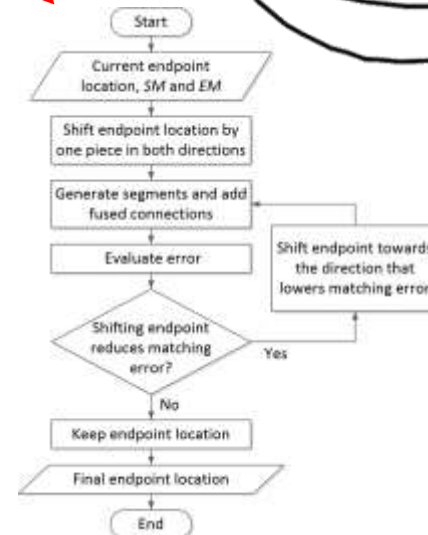
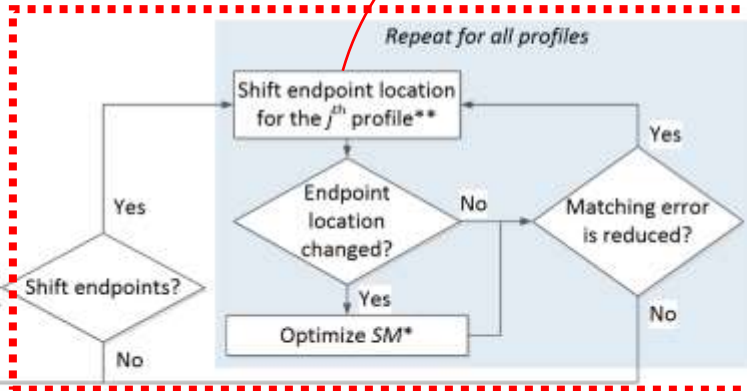
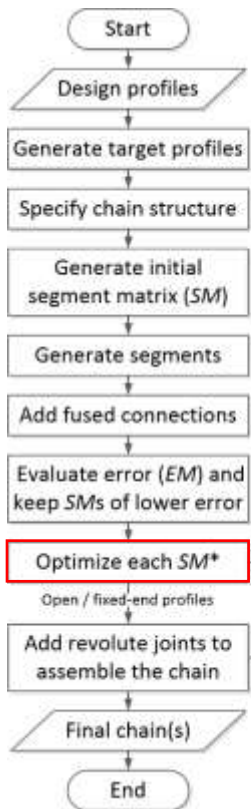


$$\mathbf{X} = \text{fmincon}(f, \mathbf{X}_0, \mathbf{C})$$

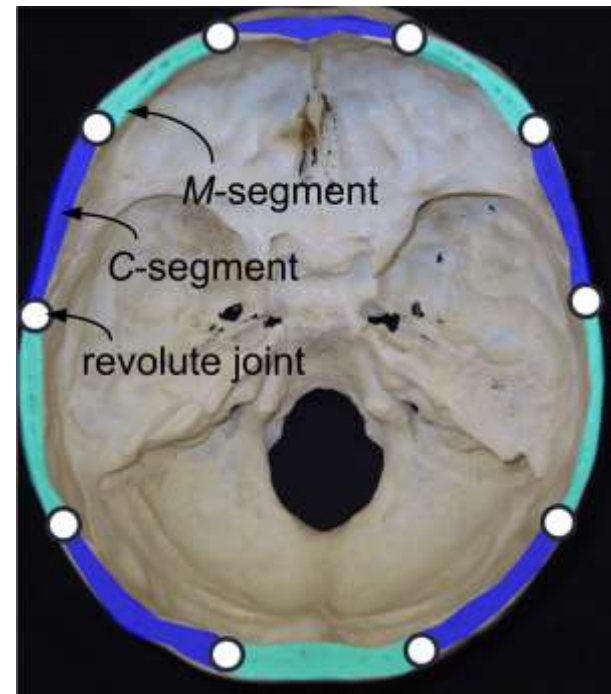
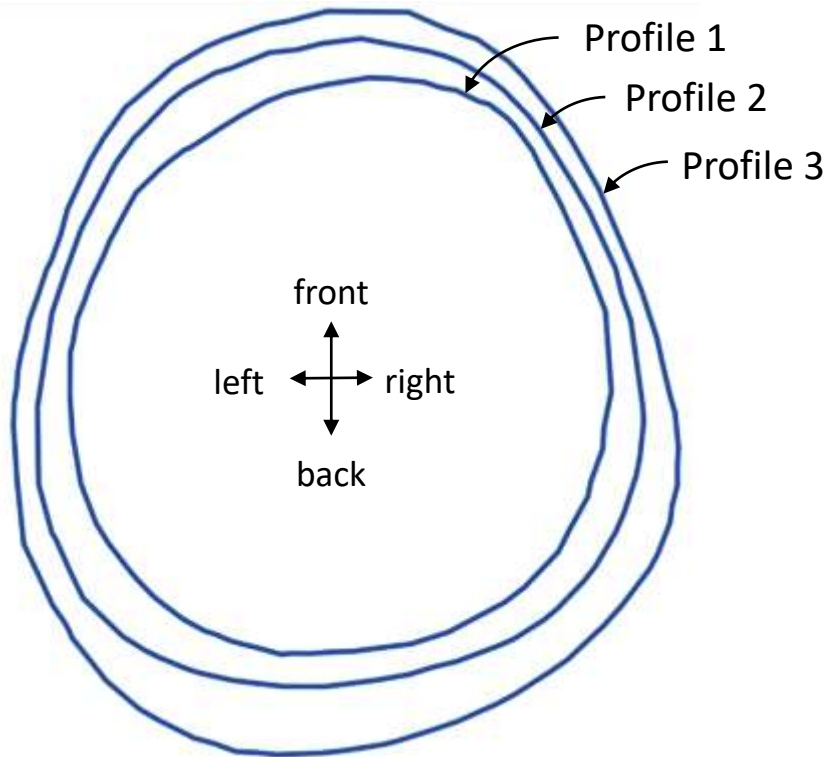
By multivariate optimization

High-Curvature Regions





Head Circumference



Head Circumference

Fused connection between endpoints

$V = [C M C M C M C M C]$

$W = [F R F R F F R F R F]$

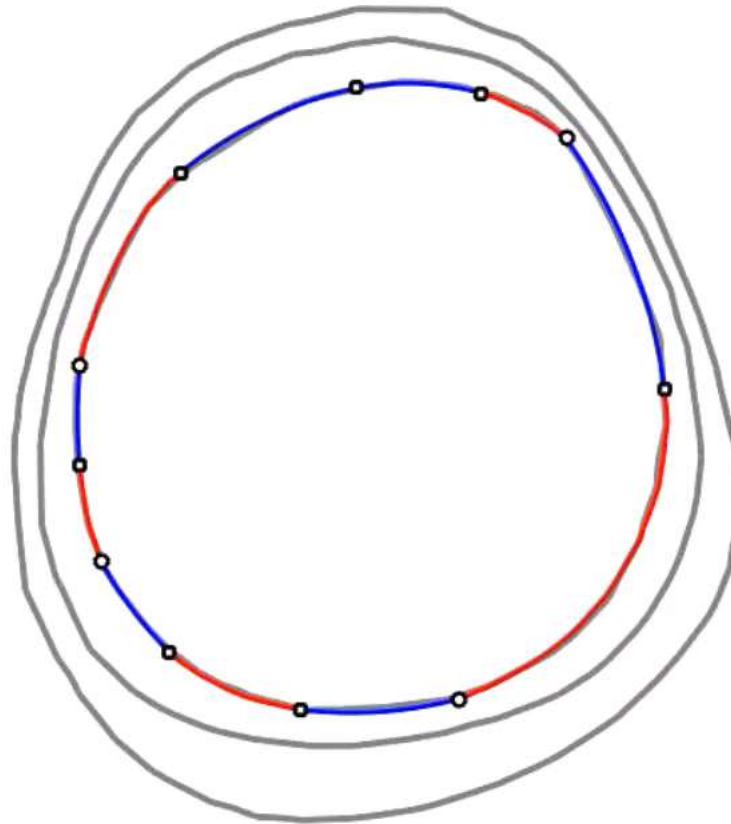
10 parameters to characterize the shape

(4 R joints & 6 C segments)

Max diameter = 877.17,

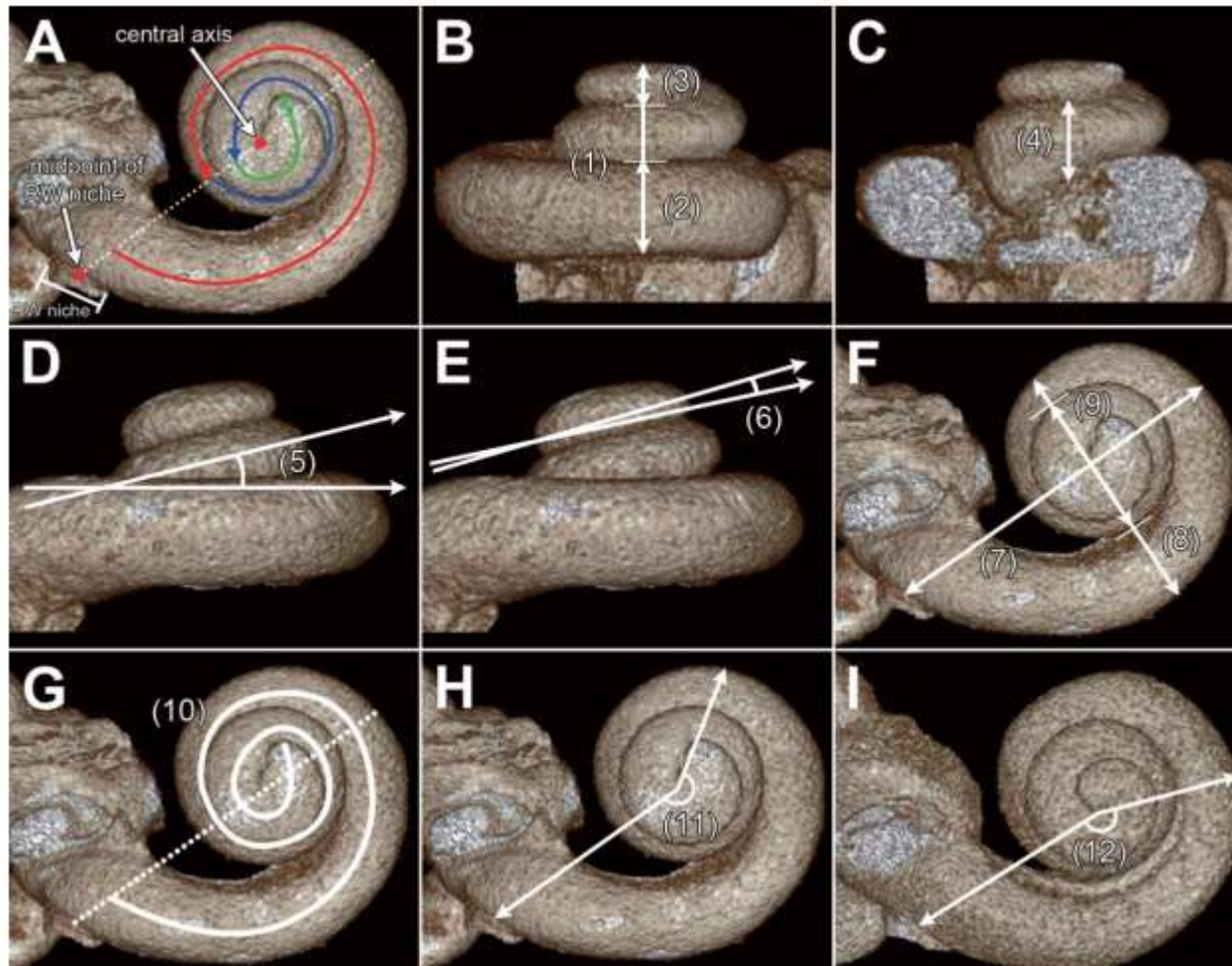
mean matching error = 6.90

(0.79% of max diameter)

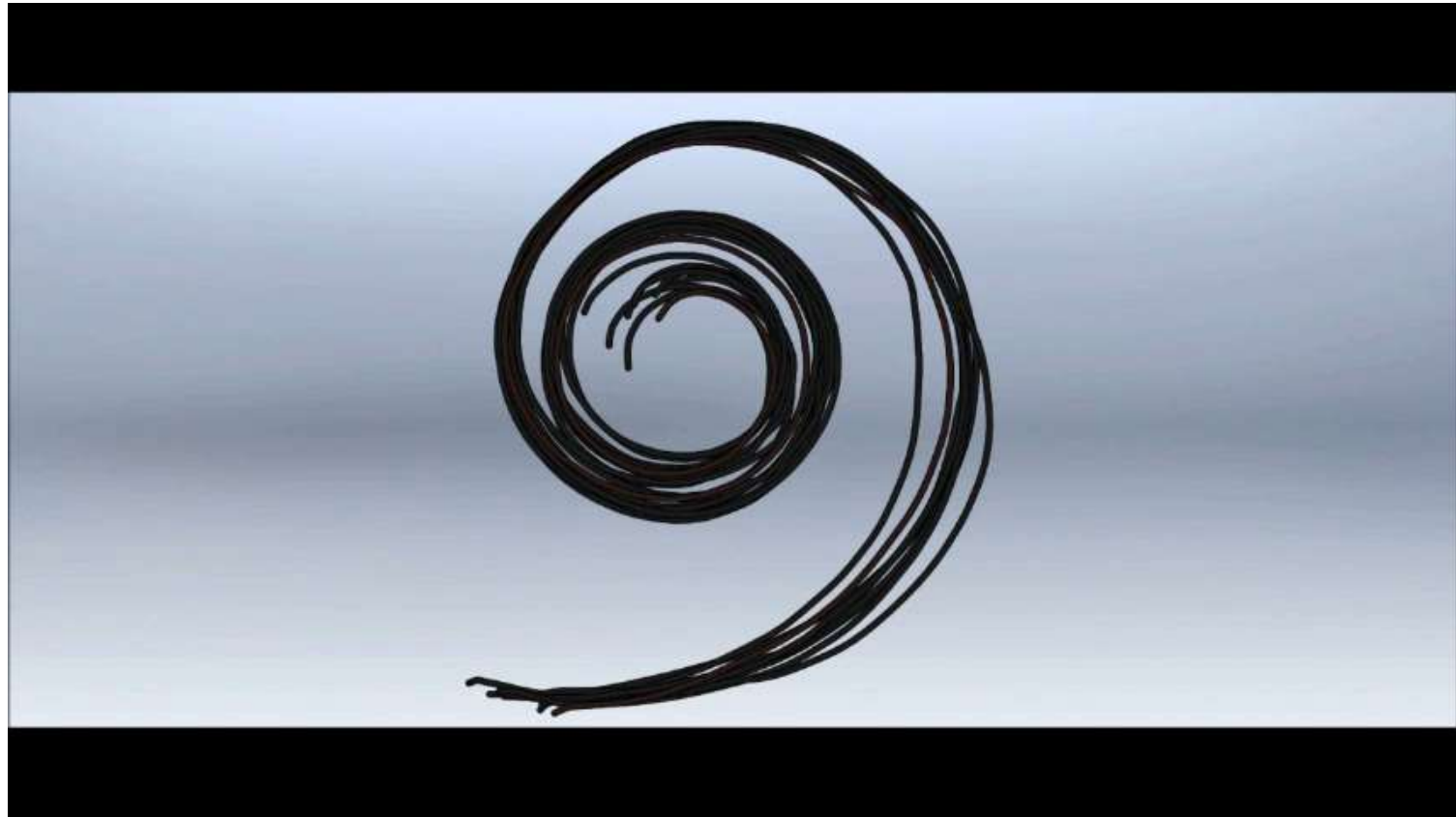


	C1	C2	C3	C4	C5	C6	R1	R2	R3	R4
Profile1	0.44	0.49	0.38	0.28	0.17	0.46	-0.27	-0.09	-0.08	-0.11
Profile 2	0.96	0.37	0.70	0.16	0.28	0.50	0.21	0.03	-0.05	0.09
Profile 3	0.68	0.73	0.43	0.62	0.46	0.38	0.16	0.26	0.02	0.09

The Cochleae

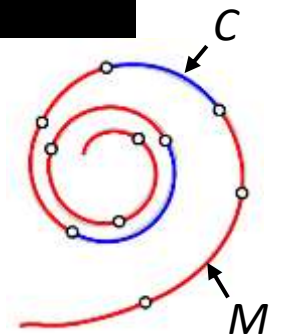


The Cochleae



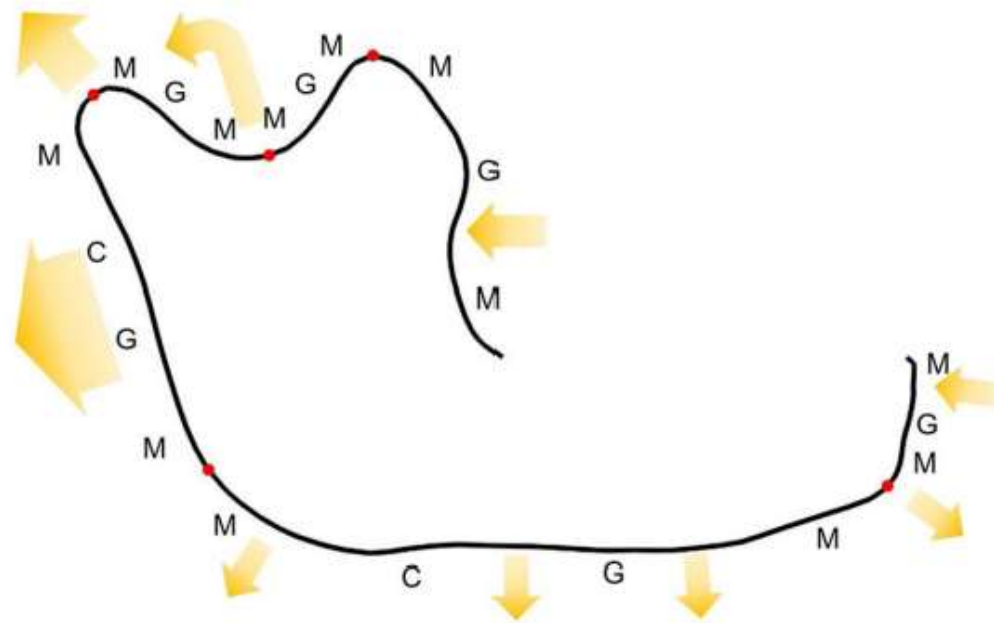
$V = [M M M M C M M C M M M]$
 $W = [R R \dots R R]$ (10 involute joints)
 12 parameters to characterize shapes
 (10 R joints + 2 C segments)

Average profile width = 798.68
 Average profile length = 976.40
 Mean matching error = 6.91



20 segments = [M G M M G M M G M M C G M M C G M M G M] and no connection
 Based on the growth pattern of the mandible

B. Li, S. Zhou, A.P. Murray, G. Subsol. "Shape-changing chains for morphometric analysis of 2D and 3D, open or closed outlines". Scientific Reports (2021 2-year impact factor: 4.380). 11, article number 21479. November 2021.



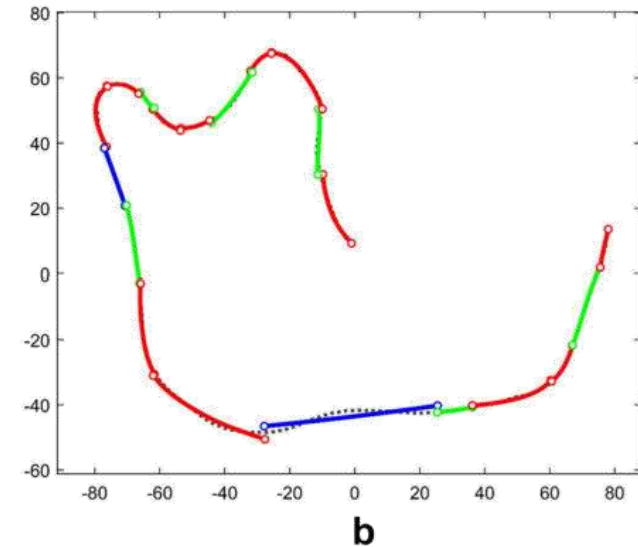
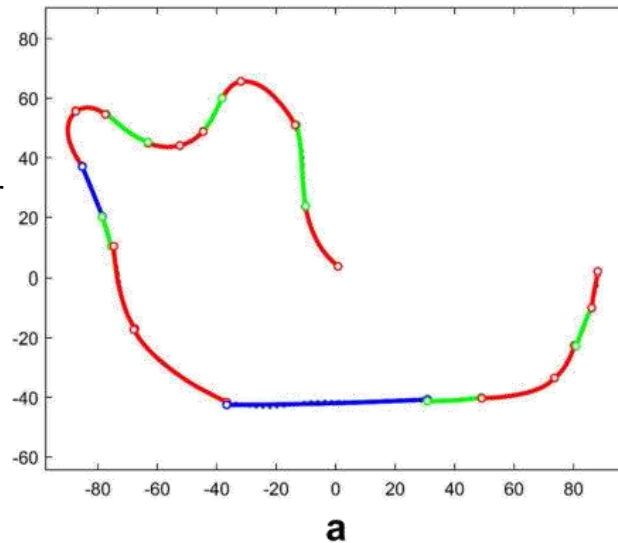


Extension 1 : Growing segment



20 segments = [M G M M G M M G M M C G M M C G M M G M] and no connection
Based on the growth pattern of the mandible

B. Li, S. Zhou, A.P. Murray, G. Subsol. "Shape-changing chains for morphometric analysis of 2D and 3D, open or closed outlines". Scientific Reports (2021 2-year impact factor: 4.380). 11, article number 21479. November 2021.



20 segments = [M G M M G M M G M M C G M M C G M M G M] but no connection
 Based on the growth pattern of the mandible

B. Li, S. Zhou, A.P. Murray, G. Subsol. "Shape-changing chains for morphometric analysis of 2D and 3D, open or closed outlines". *Scientific Reports* (2021 2-year impact factor: 4.380). 11, article number 21479. November 2021.

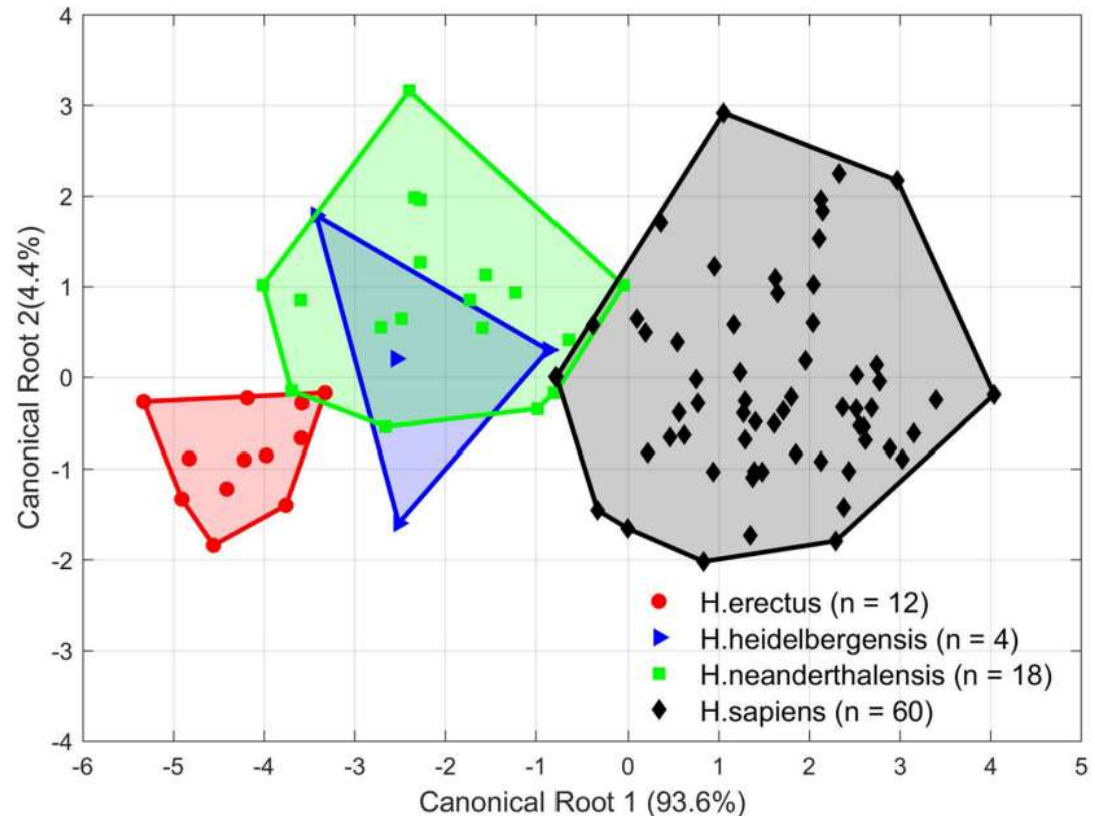


Figure 8. Canonical plot of the 94 human mandibles from four groups (*H. erectus*, *H. heidelbergensis*, *H. neanderthalensis*, and *H. sapiens*) based on the orientation changes between segments (19 variables).

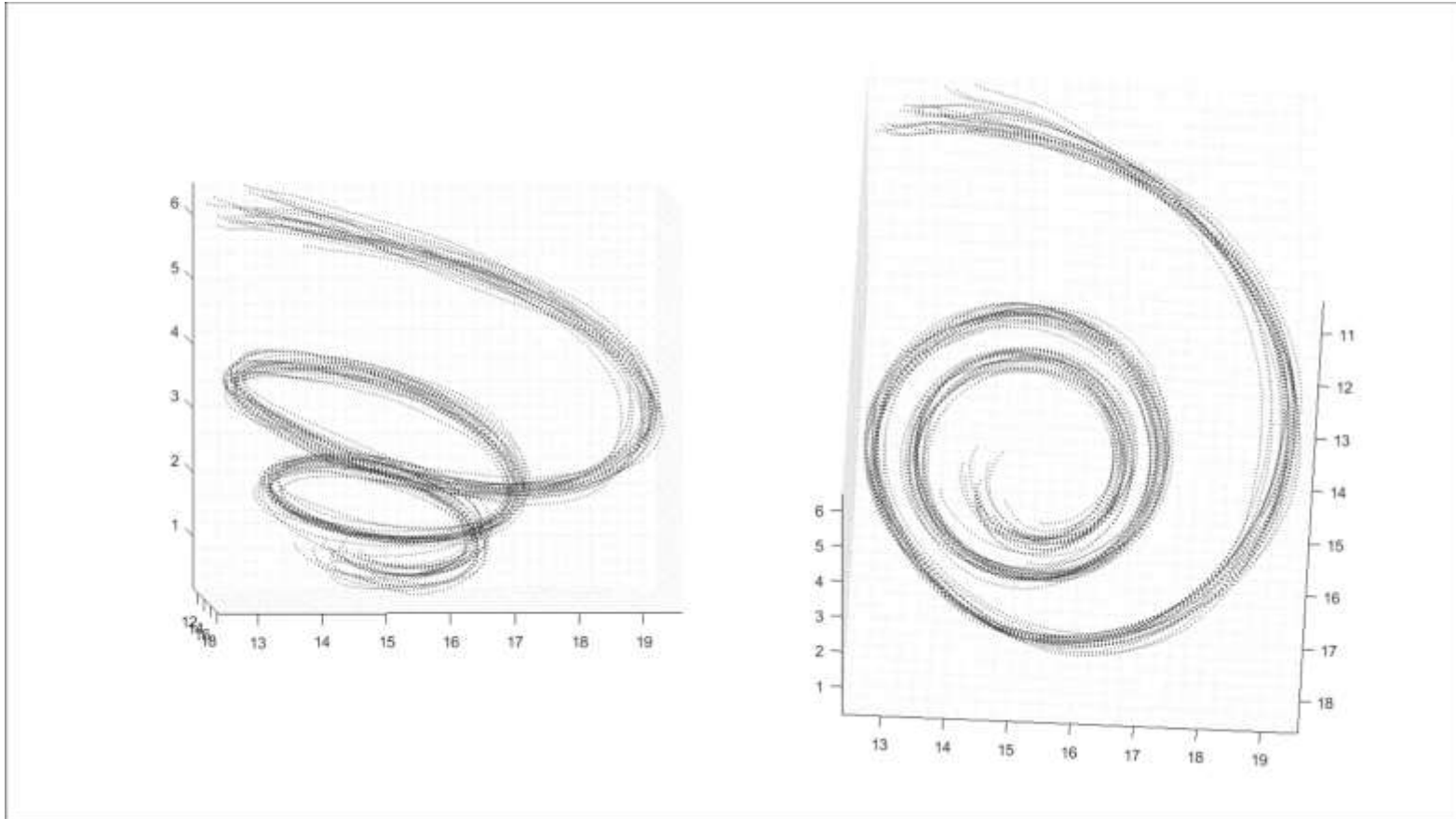


Extension 2: generalization to 3D



Examples: Cochlea in 3D

$T = [M M M M H M H M M]$ and ball joints



A new 3D morphometric method based on a combinatorial encoding of 3D point configurations: application to skull anatomy for clinical research and physical anthropology

Kevin Sol, LIRMM, Montpellier

Emeric Gioan -, ALGCo Research-Team , LIRMM, Montpellier

G rard Subsol - Research-Team ICAR, LIRMM, Montpellier

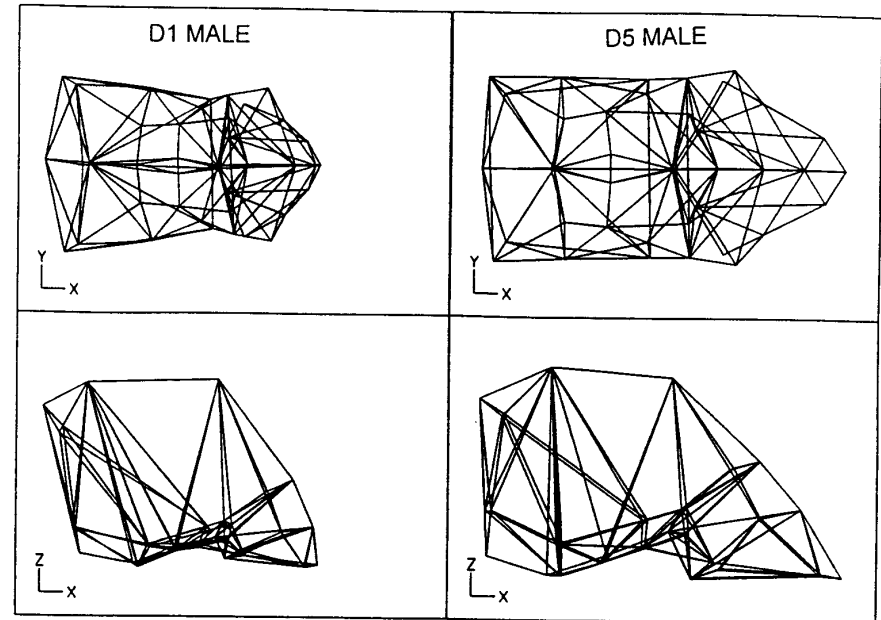
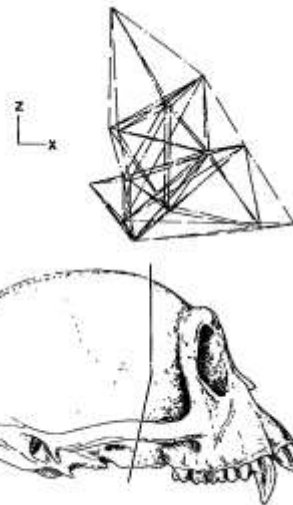
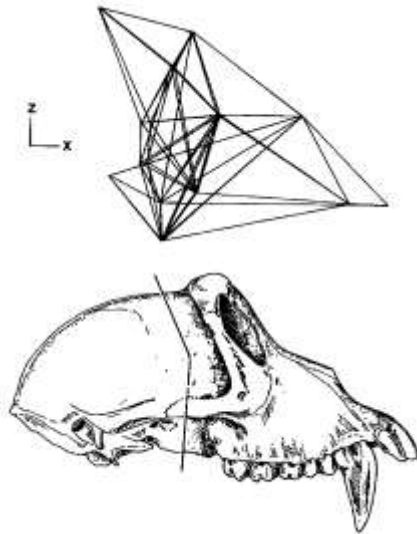
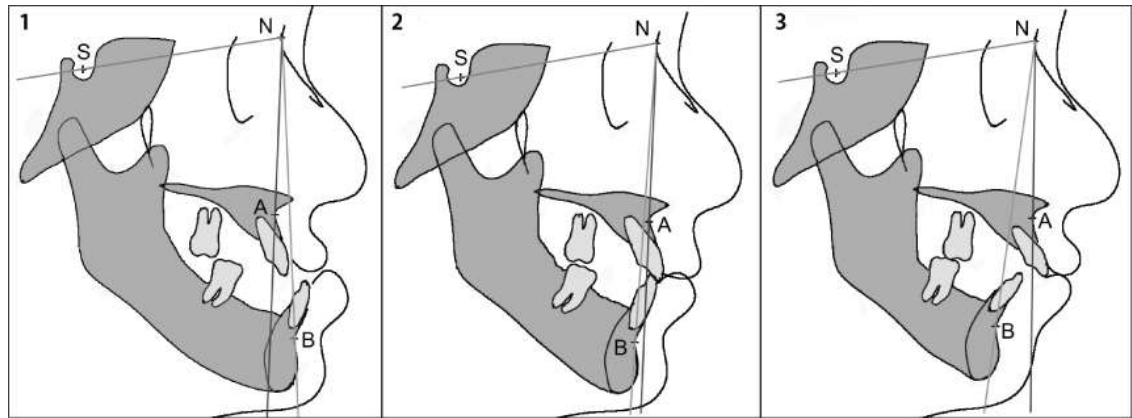
Thanks to Yann Heuz  et Joan Richstmeier - PennState Univ., U.S.A.

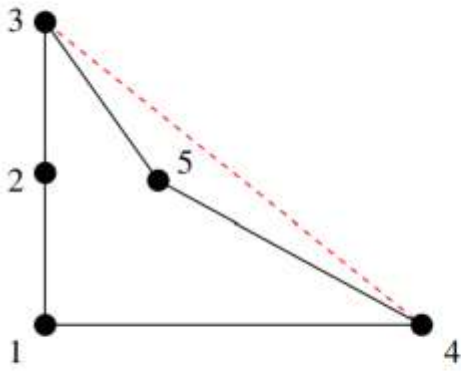
Jos  Braga et Jacques Treil – CAGT, Toulouse



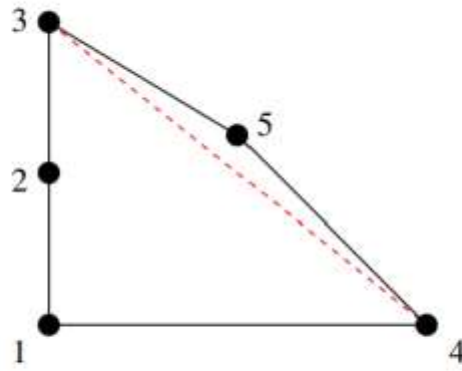
In landmark-based morphometry, shape differences may be not "metric" but rather algebraic (or topological).

Example: a subset of landmarks moves in front of an other subset (e.g. facial pathology)

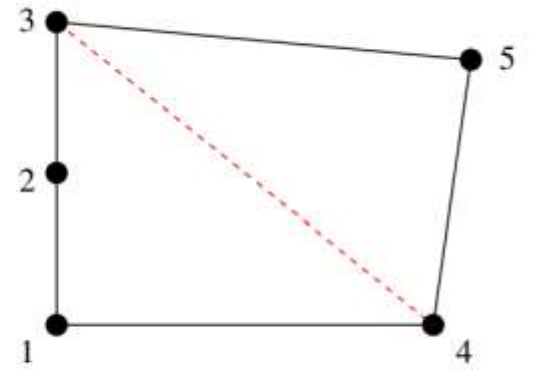




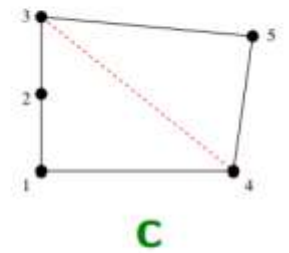
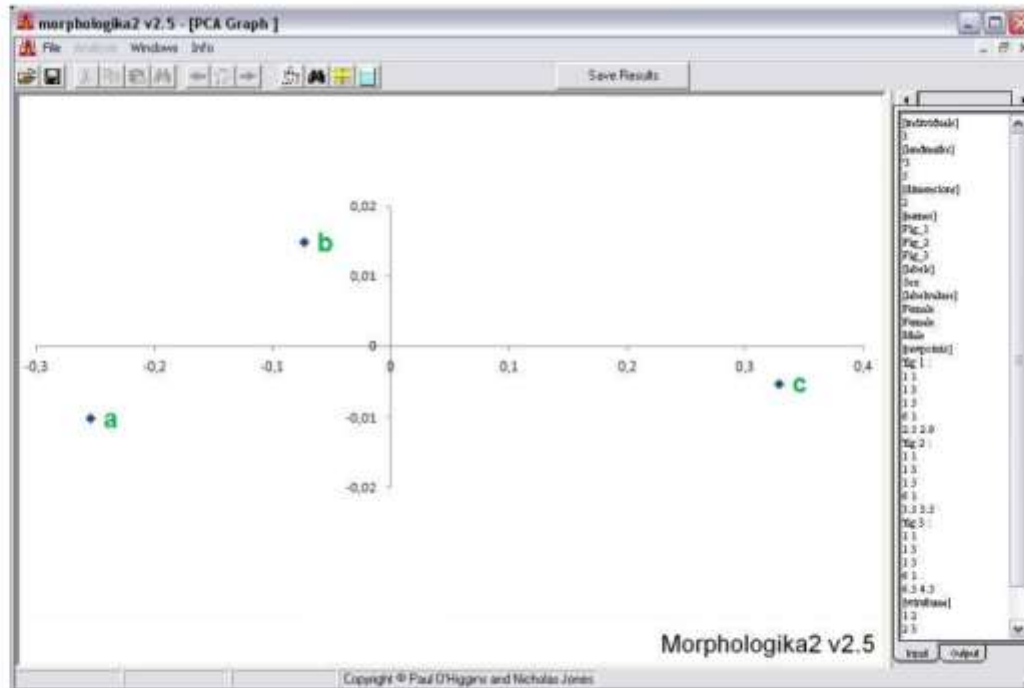
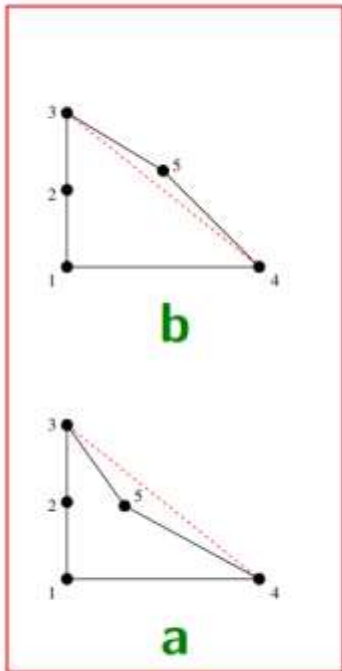
a



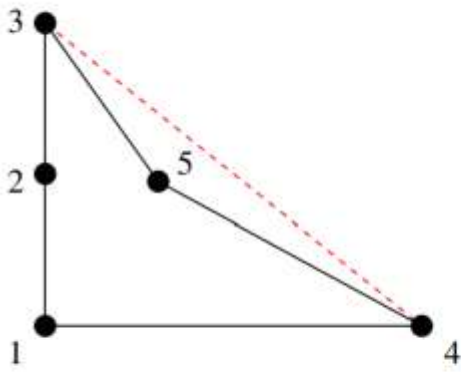
b



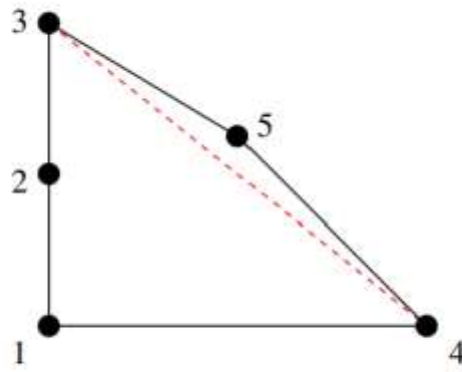
c



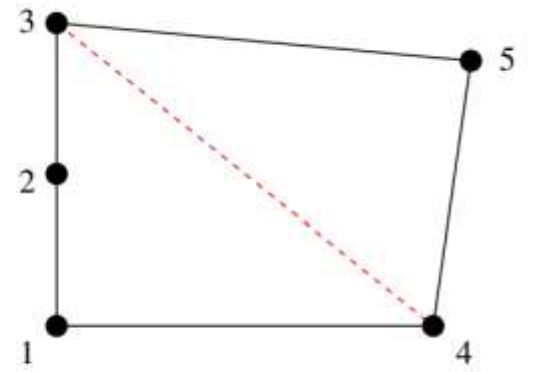
c



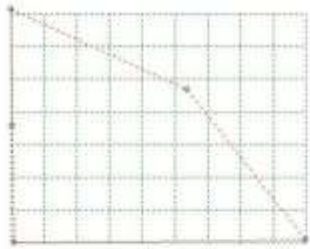
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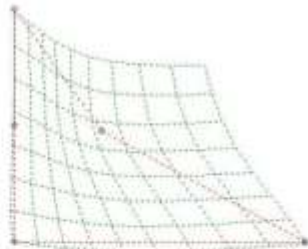
b



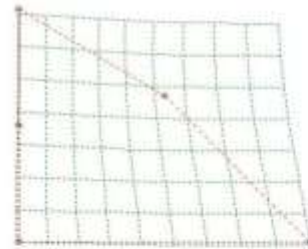
c



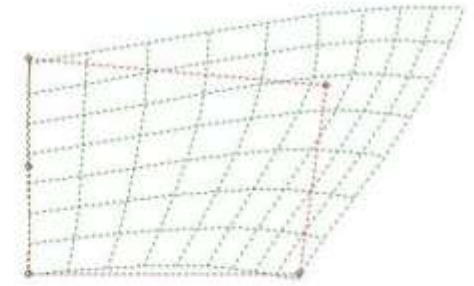
Mean
shape



a



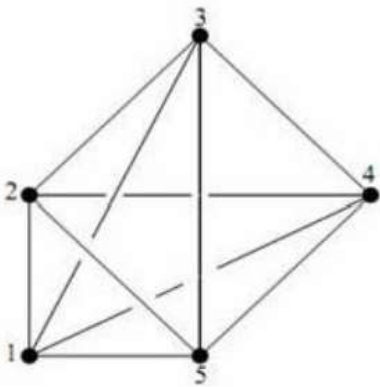
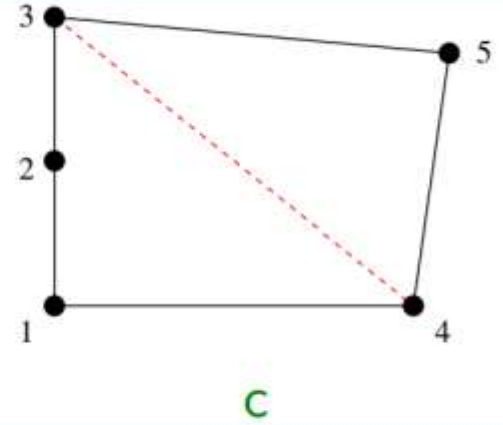
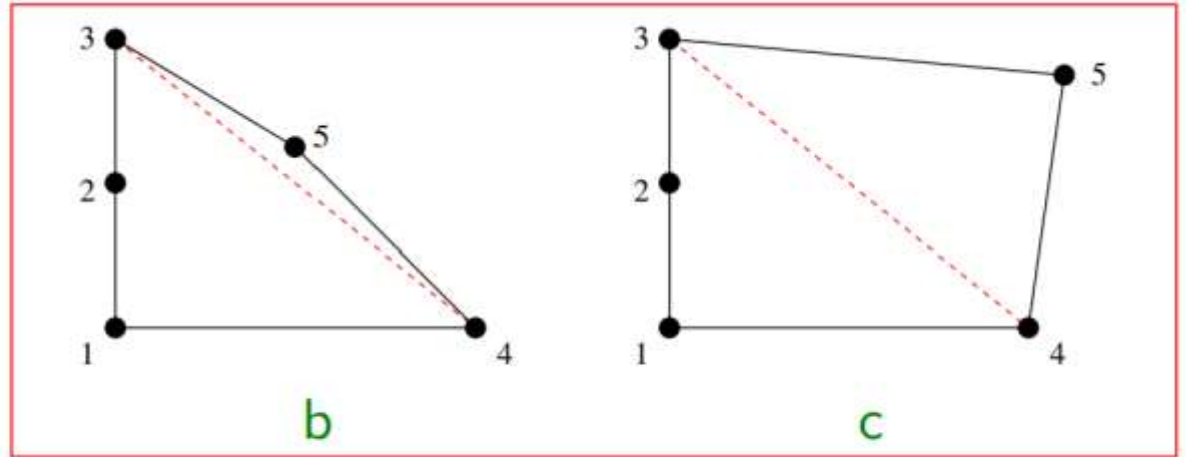
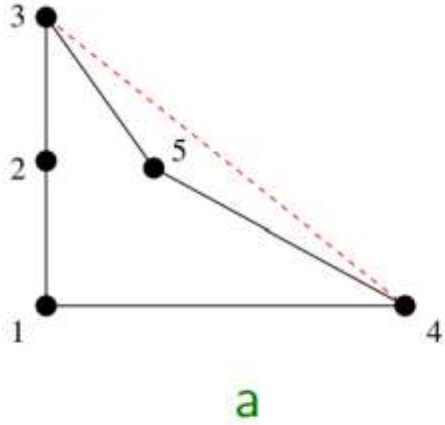
b



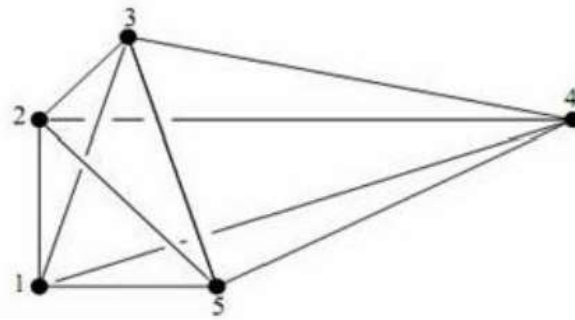
c



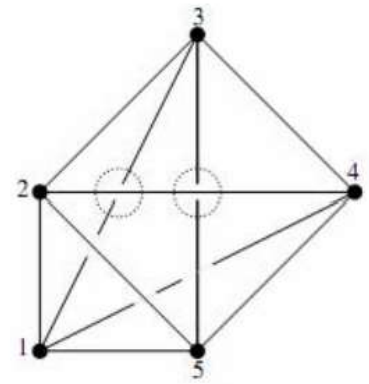
Objective



(a)



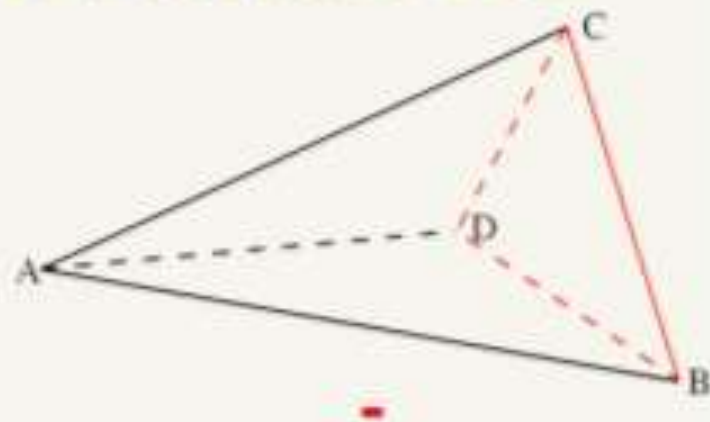
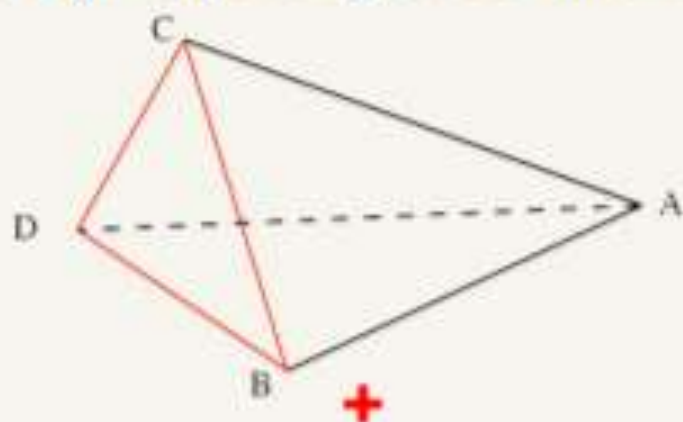
(b)



(c)

A new 3D morphometric method

In 3D, each set of 4 landmarks A, B, C, D, can be associated with a sign depending on the orientation of the tetrahedron ABCD.



For a configuration composed of n 3D landmarks, we consider

$t = \frac{n!}{(n-4)! \cdot 4!}$ tetrahedra and get a vector of t signs (+ or -) that encodes the "shape" of the configuration.

This vector defines a *combinatorial mathematical structure* called an **Oriented Matroid**.

x	y	z
17.45	26.36	13.7
14.51	26.93	11.5
15.93	26.94	12.36
17.11	26.94	13.63
18.69	26.98	14.67
19.50	27.03	15.03
15.75	29.67	10.89
14.96	29.08	11.07
...		

Landmark configuration

x	y	z
16.75	25.86	13.38
15.13	27.33	11.93
15.35	22.54	10.46
16.34	25.34	13.45
19.49	26.23	14.53
19.53	25.97	13.53
15.45	30.42	10.65
14.56	29.68	10.76
...		

...

+
-
+
+
-
+
-
+
...

Each tetrahedron gets 1 sign

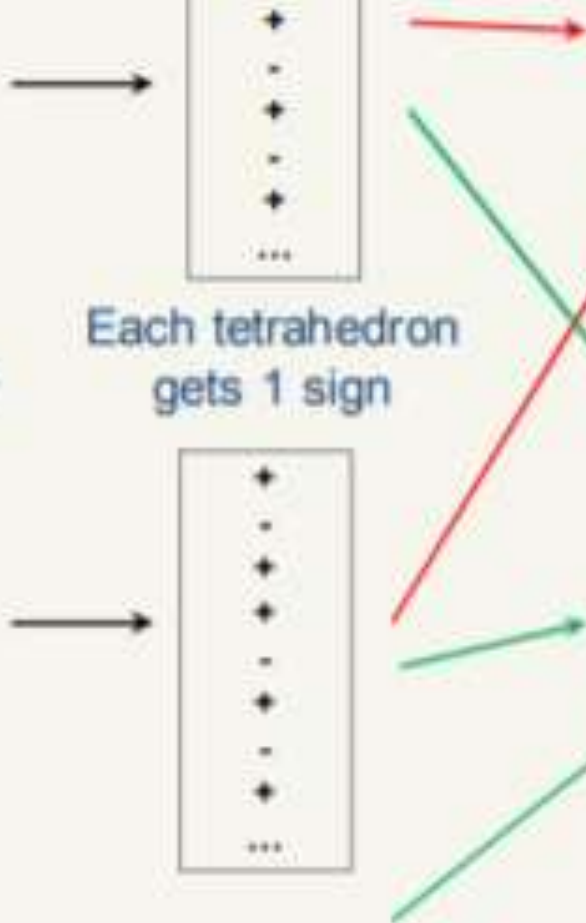
+
-
+
+
-
+
-
+
...

...

Combinatorial distance
between 2 individuals

(e.g. number of different signs)

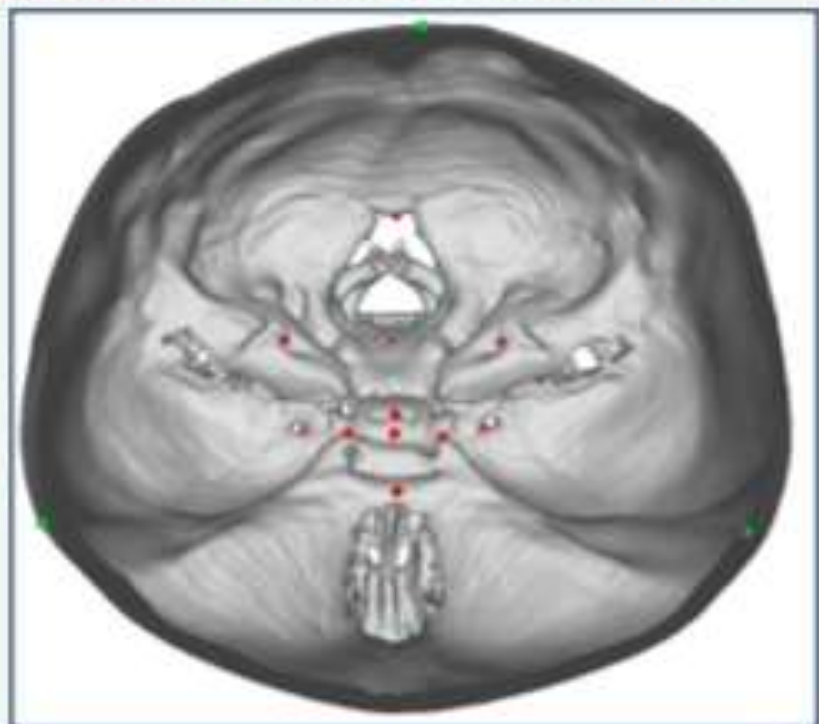
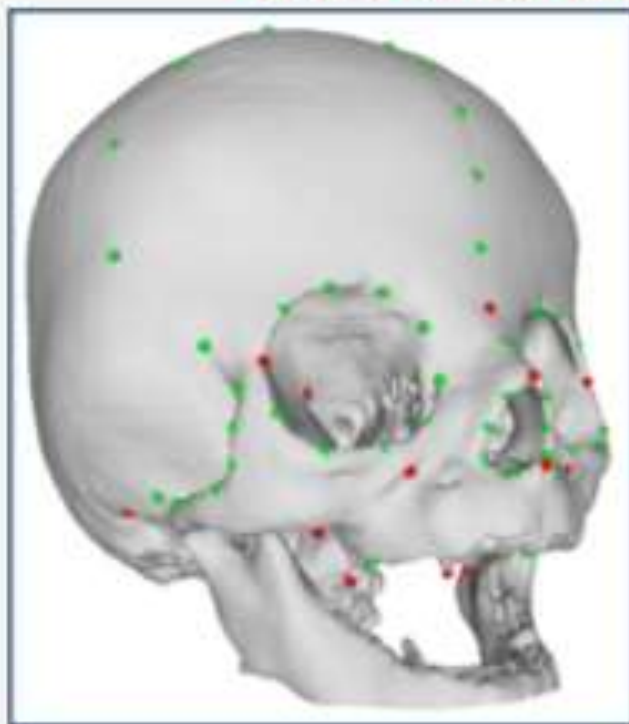
Automatic classification
of the set of all individuals



Data

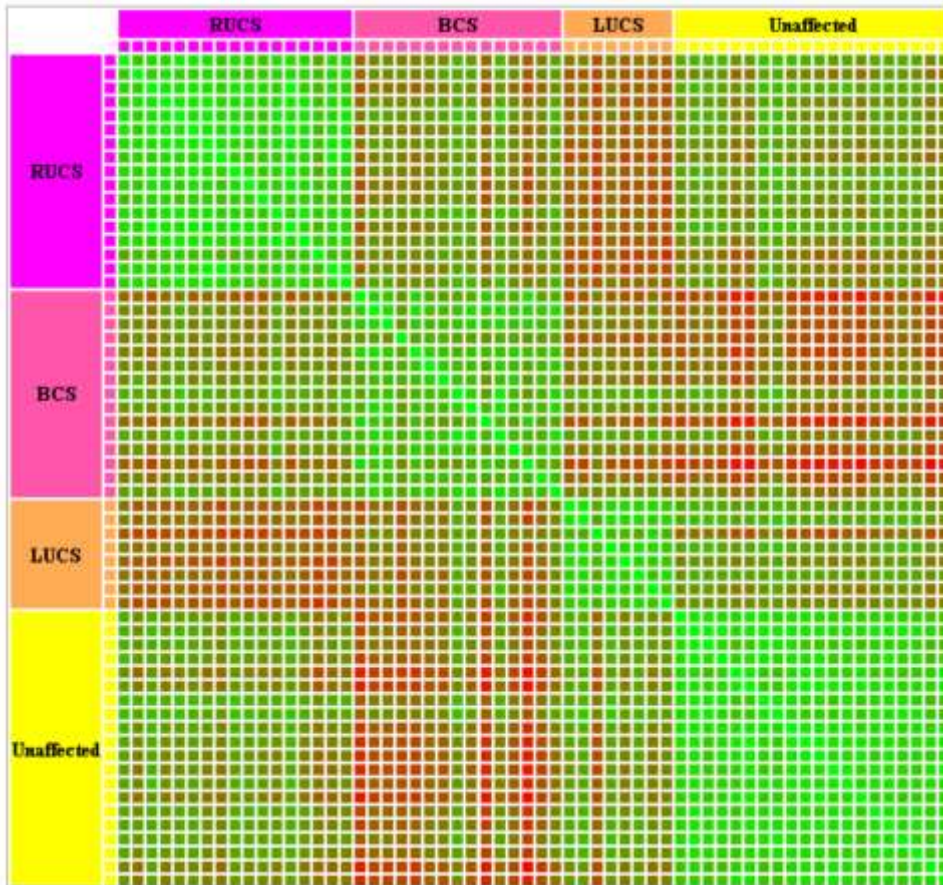
- 3D CT-images of 40 children (0.1 – 19.9 months) with craniosynostosis, i.e. premature fusion of cranial sutures
- visual evaluation and classification into 3 categories by a clinician:
 - BCS (*bicoronal*): fusions of both lateral sutures (15)
 - LUCS (*left unicoronal*): fusion of only left-side suture (8)
 - RUCS (*right unicoronal*): fusion of only right-side suture (17)
- 133 landmarks defined by an expert:

41 anatomical landmarks / **92 curve semilandmarks**.



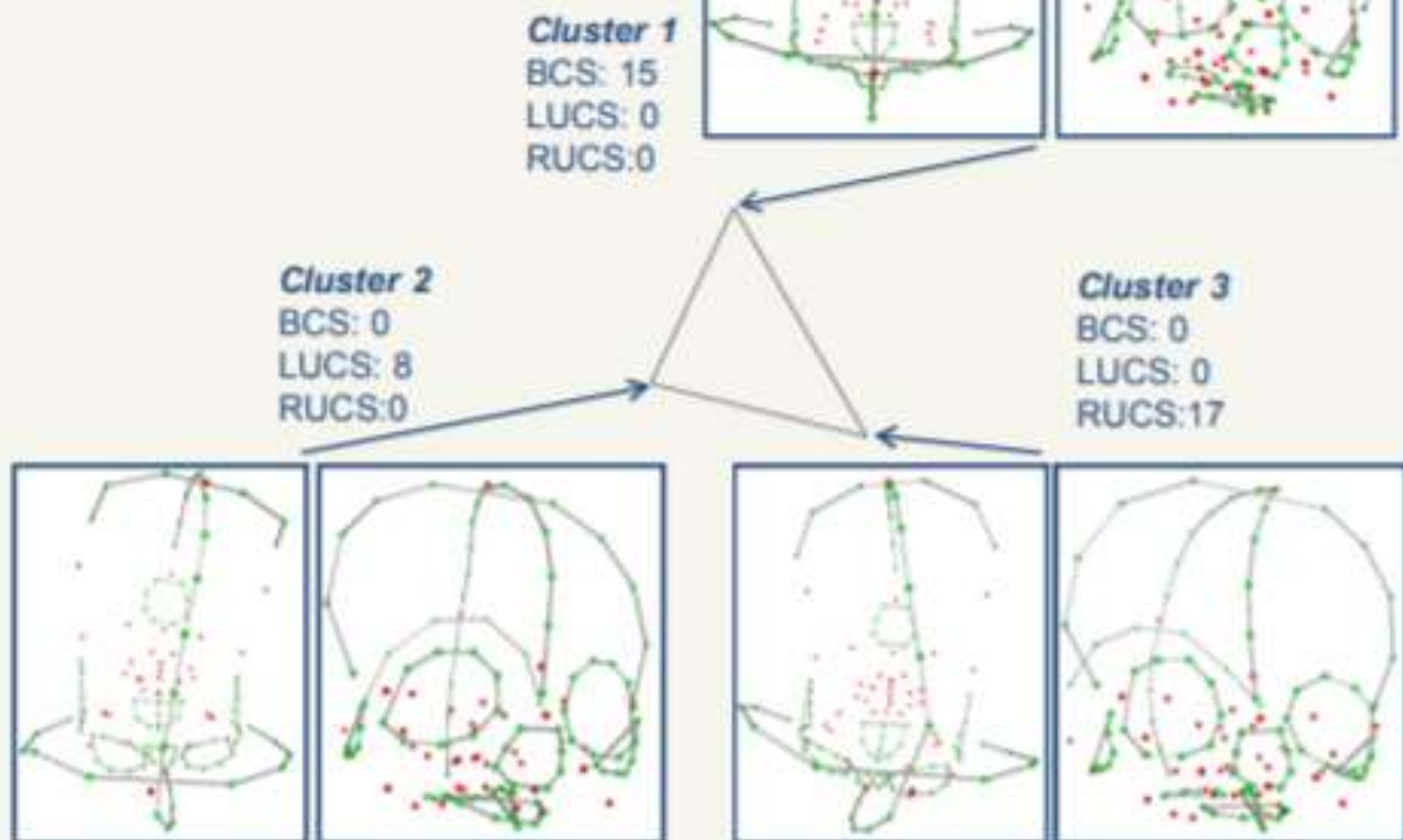
+ 20 Unaffected/Control

	RUCS	BCS	LUCS	Unaffected
RUCS	16.2%	27.6%	34%	25.9%
BCS	27.6%	20.2%	30%	33.4%
LUCS	34%	30%	17.5%	26%
Unaffected	25.9%	33.4%	26%	15.8%



Automatic classification

Cluster analysis by K-means criterion / algorithm using this combinatorial distance.



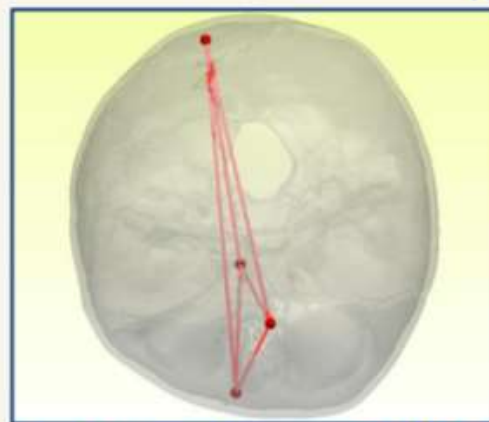
Mathematically, K-means clustering based on this combinatorial distance provides classification matching diagnostic categories.

E. Gioan, K. Sol, G. Subsol. "A combinatorial method for 3D landmark-based morphometry: application to the study of coronal craniosynostosis". 15th International Conference on Medical Image Computing and Computer Assisted Intervention, Nice (France), October 2012, Lecture Notes in Computer Science 7512, p. 533–541, Springer, 2012.

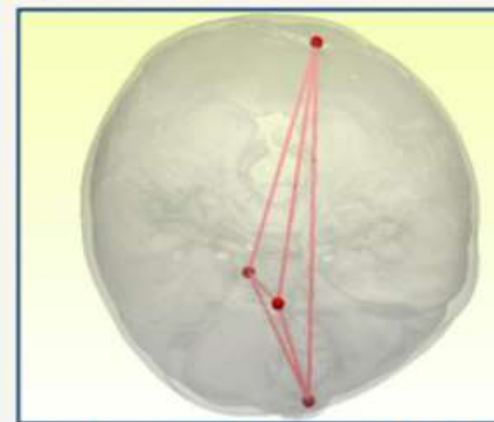
K. Sol. "Une approche combinatoire novatrice fondée sur les matroïdes orientés pour la caractérisation de la morphologie 3D des structures anatomiques". Ph.D. Thesis in Computer Science, University of Montpellier II (France), December 2013.

Some Characterizations of Classes

(using only the 41 anatomical landmarks)

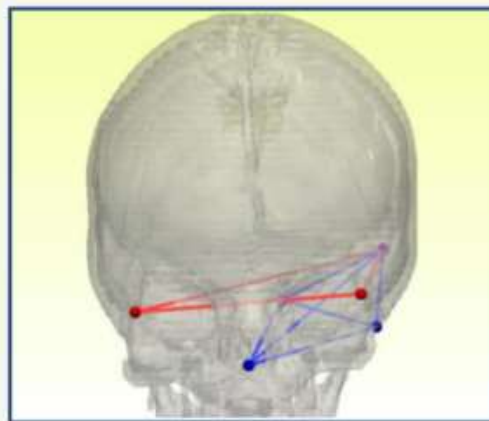


$$[\chi(b_1) = +1] \Leftrightarrow \text{RUCS}$$

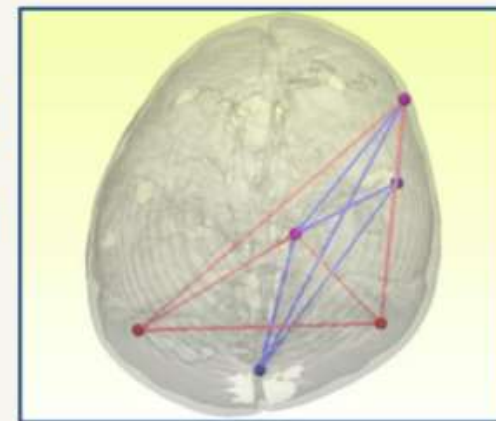


$$[\chi(b_2) = +1] \Leftrightarrow \text{LUCS}$$

- RUCS and LUCS are characterized by the sign of only 1 basis.
- The 2 basis b_1 and b_2 are symmetric w.r.t. the median sagittal plan.



$$[\chi(b_3) = -1] \\ \text{and} \\ [\chi(b_4) = -1] \\ \Leftrightarrow \\ \text{BCS}$$



- The signs of 2 bases characterize the category **BCS**.
- Based on the discriminability, we found a subset \mathcal{J} of 5 bases and a vector x in $\{-1, 1\}^{\mathcal{B}}$ such as: M is **unaffected** if and only if $M \in B(\mathcal{J}, x, 2)$ (i.e. the signs of at least 3 of these 5 bases are the same in x and χ_M).

