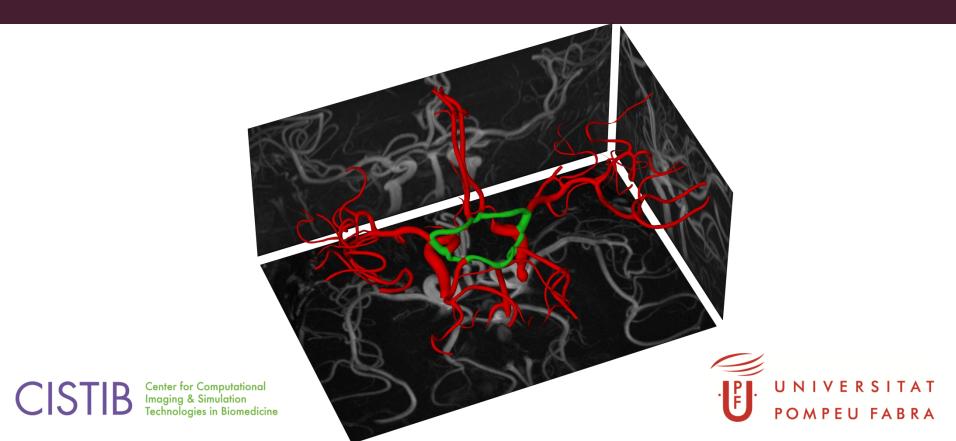
Geometric Modeling and Characterization of the Circle of Willis

Hrvoje Bogunović

CCVW 2012 / Zagreb, Croatia / 2012-09-21



Clinical motivation

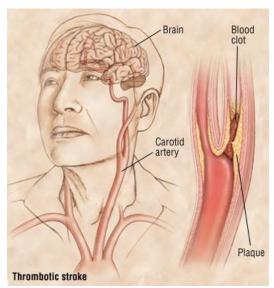
- Brain depends critically on its blood supply
 - 15% of cardiac output
- Cerebrovascular disease can lead to a stroke
 - Atherosclerosis
 - Aneurysms
- Stroke
 - Major cause of disability
 - Second leading cause of death
 - Western society
 - Ischemic (80%)
 - Thrombotic, embolic
 - Hemorrhagic (20%)

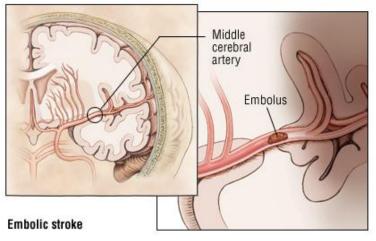






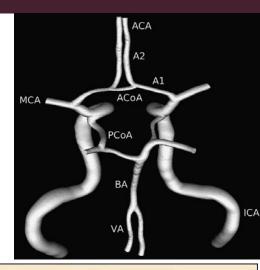


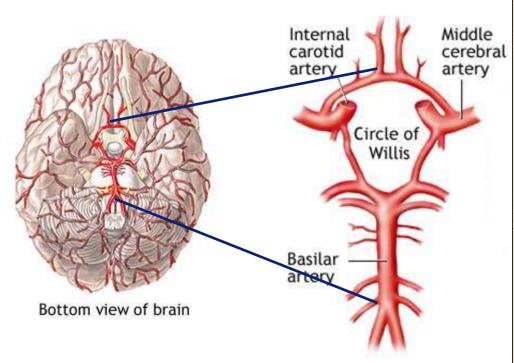


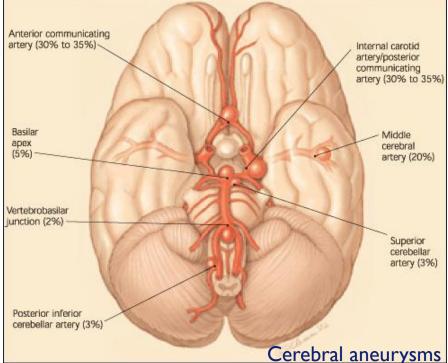


Clinical motivation

- Focus on the Circle of Willis
 - Conducts the blood flow from the anterior-left,
 anterior-right and posterior circulation to the brain
 - Arterial system forms a cycle by design
 - Common site of pathologies
 - Large anatomical variability



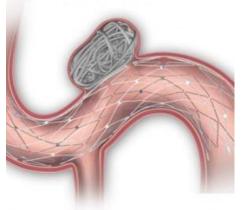


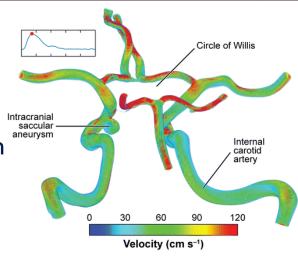


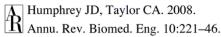
Role of vascular geometry

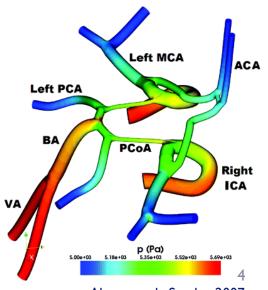
- Geometric risk factors*
 - Cerebral aneurysms occur at specific locations, at or near high curvature
 - Atherosclerosis often occurs at carotid bifurcation or close to arterial bendings
 - Hypothesis: hemodynamics responsible
 - Dependent on geometry

- Endovascular treatment difficulty
 - Device selection, and intervention planning
 - Stent fitness









Alnaes et al., Stroke, 2007.

^{*} Friedman et al., Atherosclerosis, vol. 46, pp. 225–231, Feb. 1983.

Technical challenges

Characterization

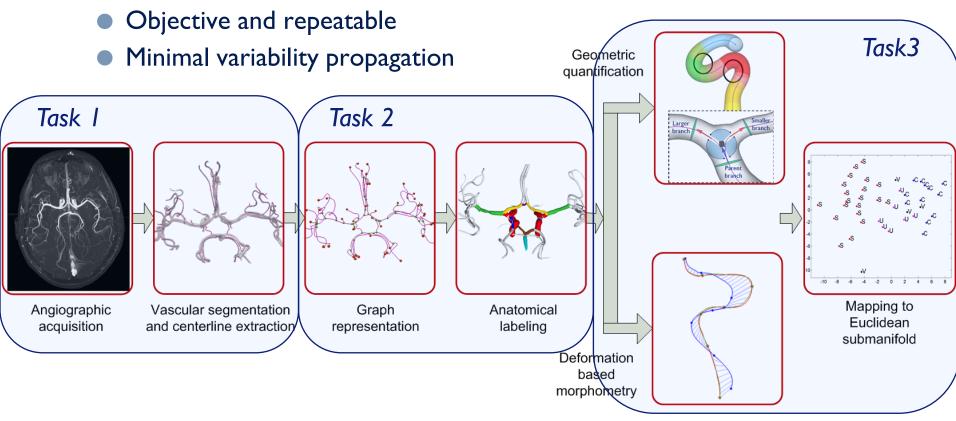
Statistical

Analysis

- Processing pipeline for image-based, subject-specific geometry characterization of the Circle of Willis
- High level of automation

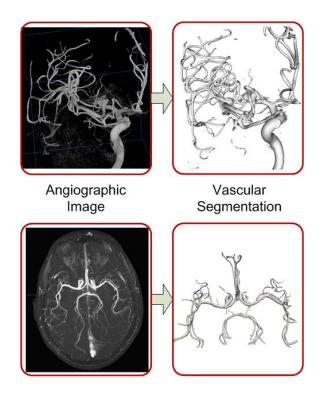
Geometric Modeling

Image Acquisition



Anatomical Labeling

TI: SEGMENTATION OF CEREBRAL VASCULATURE



H. Bogunović, J.M. Pozo, M.C. Villa-Uriol, C.B.L.M. Majoie, R. Van Den Berg, H.A.F. Gratama Van Andel, J.M. Macho, J. Blasco, L. San Román, A.F. Frangi: Automated segmentation of cerebral vasculature with aneurysms in 3DRA and TOF-MRA using geodesic active regions: an evaluation study. *Medical Physics*, vol. 38(1), Jan. 2011, pp. 210-222.

TI: Segmentation

Introduction

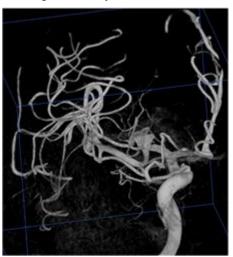
- Objective: From angiographic image obtain accurate geometric model of vasculature containing pathologies (aneurysms)
 - Focus on 3DRA and MRA imaging modalities

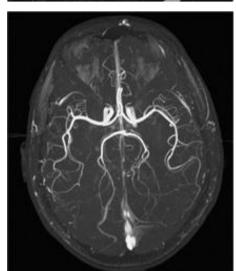
■ 3DRA

- Invasive: contrast injected into artery + radiation
- Treatment planning and evaluation
- High image spatial resolution (0.2x0.2x0.2 mm³)

■ ToF-MRA

- Non-invasive, no contrast, no radiation
- Screening and follow-up. Volunteers.
- Anisotropic spatial resolution $(0.5 \times 0.5 \times 0.8 \text{ mm}^3)$
- Full CoW observed





TI: Segmentation

Methodology

 $\Psi(x, t)$

-2.4|-1.3|-0.6|-0.7|-0.8|-1.8

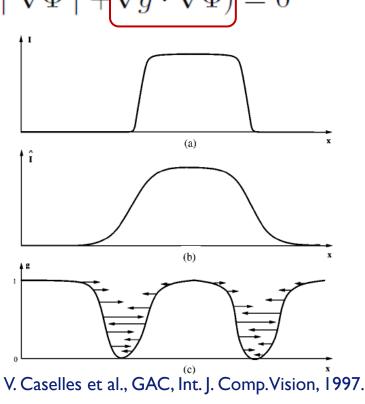
-1.6 - 0.6 0.4 -0.7 -0.6 -1.6

Geodesic Active Regions (GAR)

- Geometric deformable model using the level set framework
 - Completely automated (including initalization)

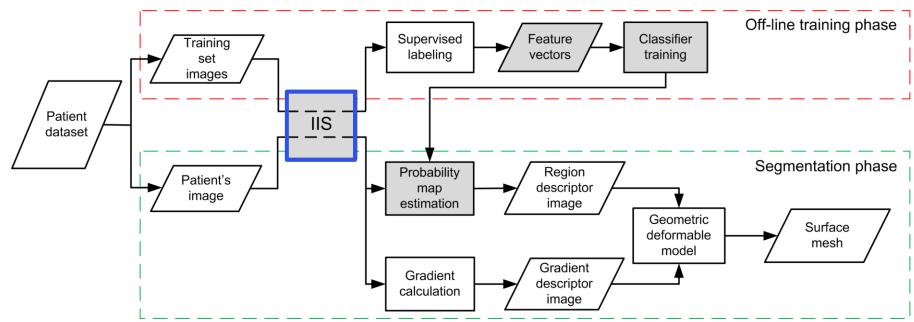
$$\frac{\partial}{\partial t} \Phi + \zeta (k_{out} - k_{in}) | \nabla \Phi | - \eta (\varepsilon g K_m) | \nabla \Phi | + \nabla g \cdot \nabla \Phi) = 0$$

- Gradient term:
 - From geodesic active contour (GAC)
 - Surface at local gradient maximum (objective) $g = \frac{1}{1 + |\nabla \hat{I}|^p}$
- Region term: use of training-set
 - Learns vessel and background features
 - Differential invariants (multiscale)
 - Intensity
 - Gradient magnitude, Laplacian ...



GAR workflow

- 3DRA and MRA: No correspondence tissue-intensity
- Introduction of the image intensity standardization (IIS)



M. Hernandez and A. F. Frangi, Medical Image Analysis, 2007.

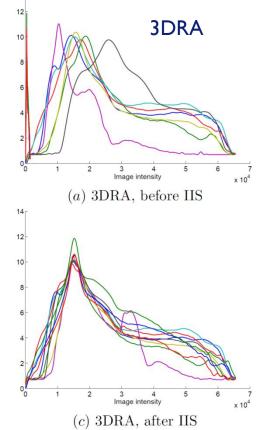
H. Bogunović, et al., SPIE Medical Imaging, 2008.

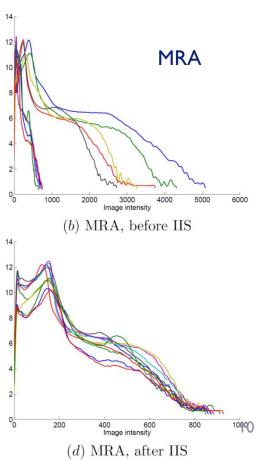
H. Bogunović et al, Medical Physics, 2011.

TI: Segmentation

GAR - Image Intensity Standardization (IIS)

- Non-linear signal registration to a histogram reference
 - Similarity metric: normalized cross-correlation
 - Reference per modality



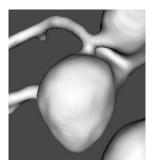


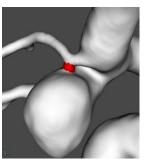
TI: Segmentation

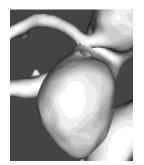
Summary

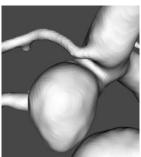


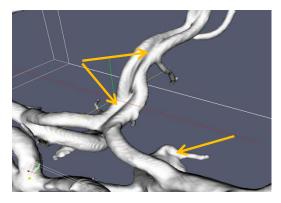
- Difficult problem:Touching vessels
 - Requires manual post-processing





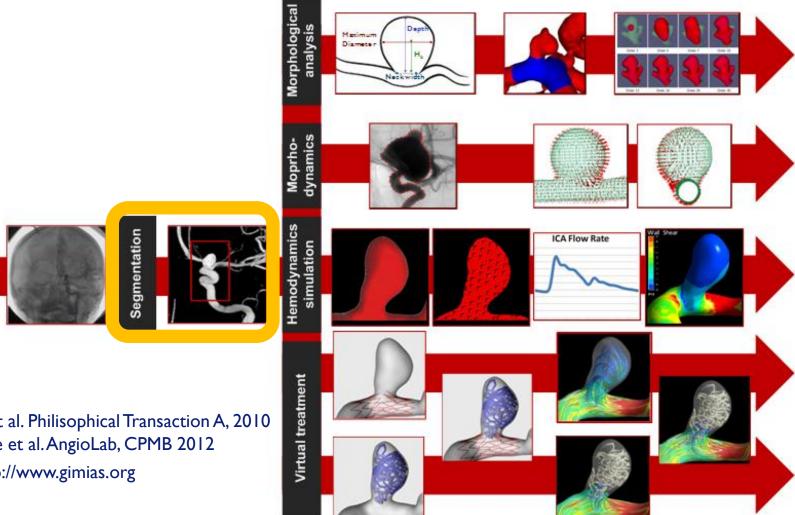






Summary

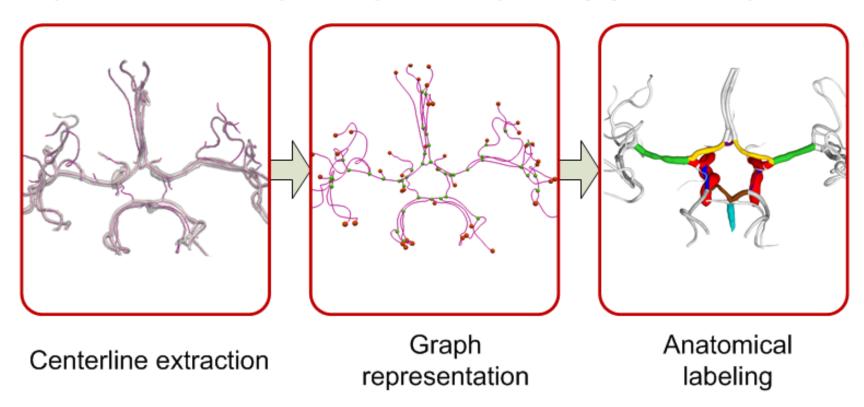
Use in aneurysm management pipeline



M-C.Villa-Uriol et al. Philisophical Transaction A, 2010 I. Larrabide et al. AngioLab, CPMB 2012 http://www.gimias.org

Acquisition

T2: ANATOMICAL LABELING OF THE CIRCLE OF WILLIS

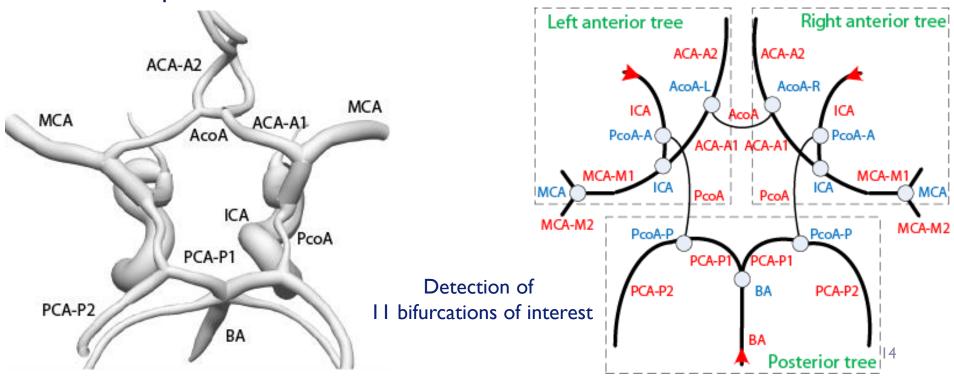


H. Bogunović, J.M. Pozo, R. Cardenes, L. San Roman, and A.F. Frangi: Anatomical labeling of the Circle of Willis using maximum a posteriori estimation. *Under Review.*

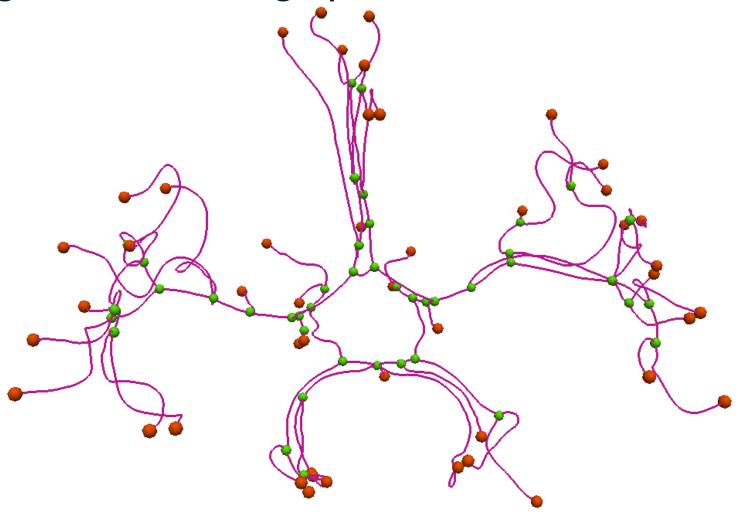
T2:Anatom. Labeling

Introduction

- Objective: Establish anatomical correspondence between subjects in the region of the Circle of Willis (CoW)
 - Enables subsequent geometric characterization and population analysis of the entire CoW
- Arteries form a graph containing a cycle with multiple roots
 - Unique to the CoW



Segmentation and graph extraction

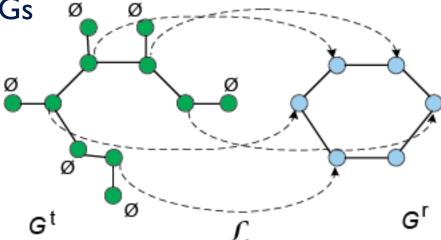


T2: Anatom. Labeling

Methodology

Labeling as mapping between ARGs

$$\mathcal{L}: V^t \to V^r \cup \{\emptyset\}$$



- Availability of a knowledge base (KB)
 - Set of reference graphs $\{V^r\}$
 - Sample vertex attributes $\{A^r\}$
 - Joint Bol configuration appearance $\{V^r\}$
- Maximum a posteriori probability $\mathcal{L}^* = \arg \max_{\mathcal{L}} P(\mathcal{L}|\hat{G}^t, KB)$

$$P(\mathcal{L}|G^{t}, A^{t}, R^{t}, KB) \propto p(A^{t}|\mathcal{L}, G^{t}, R^{t}, KB)P(\mathcal{L}|G^{t}, R^{t}, KB)$$

likelihood

prior

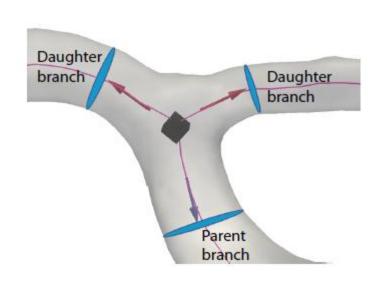
Likelihood term $p(A^{t}|\mathcal{L}, G^{t}, R^{t}, KB)$

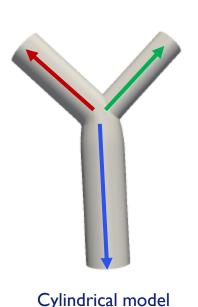
■ Independency of attributes assumed

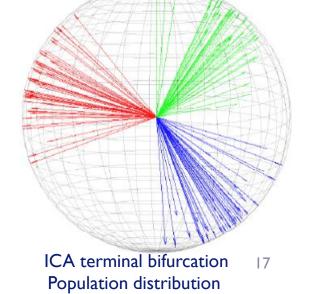
$$p(A^{t}|\mathcal{L}, G^{t}, R^{t}) = p(A^{t}|\mathcal{L}, V^{t}) = \prod_{i=1} p(a_{i}^{t}|\mathcal{L}(v_{i}^{t}))$$

- Each bifurcation a 7-tuple.
 - Element of Riemannian manifold

$$a = (\mathbf{x}, \mathbf{n}_0, r_0, \mathbf{n}_1, r_1, \mathbf{n}_2, r_2) \in \mathcal{M} = \mathbb{R}^3 \times S^2 \times \mathbb{R}^+ \times S^2 \times \mathbb{R}^+ \times S^2 \times \mathbb{R}^+$$





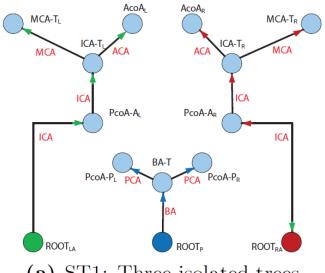


Prior term $P(\mathcal{L}|G^{t}, R^{t}, KB)$

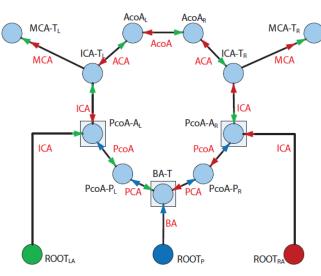
- Assures the labels follow the ordering of the reference graph
 - Only compatible labelings (L_c^t) allowed

$$P(\mathcal{L}|G^{\mathbf{t}}, R^{\mathbf{t}}) = \begin{cases} 0 & \text{if } \mathcal{L} \notin L_{\mathbf{c}}^{\mathbf{t}} \\ P(V_{\mathcal{L}}^{\mathbf{r}}) & \text{if } \mathcal{L} \in L_{\mathbf{c}}^{\mathbf{t}}; \end{cases}$$

 \blacksquare L_c : Labelings compatible with a reference graph



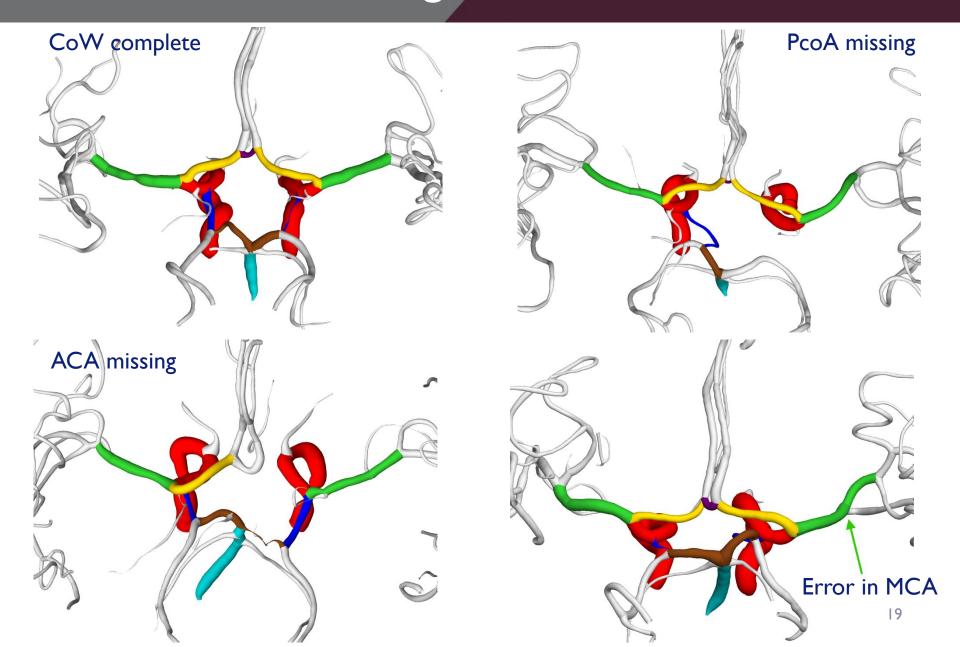
(a) ST1: Three isolated trees



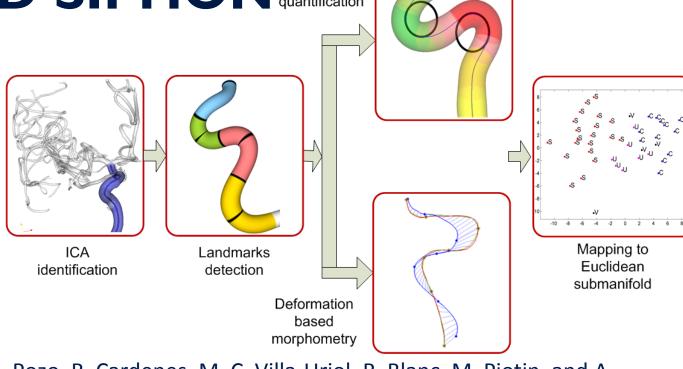
(h) ST8: Complete CoW

T2: Anatom. Labeling

Results



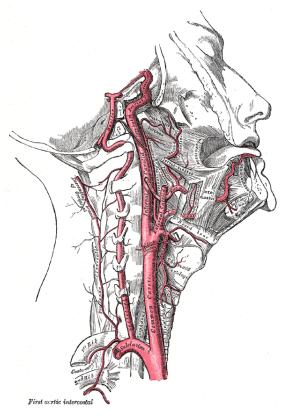
T3: GEOMETRIC CHARACTERIZATION OF THE CAROTID SIPHON Geometric quantification

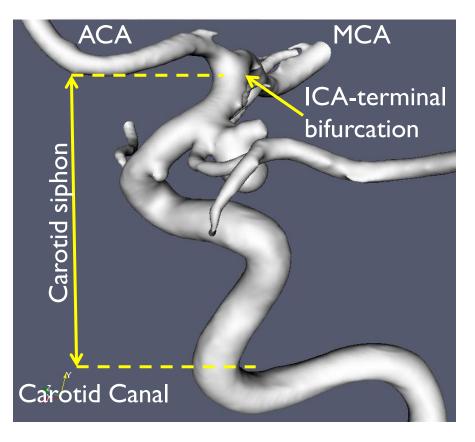


H. Bogunović, J. M. Pozo, R. Cardenes, M. C. Villa-Uriol, R. Blanc, M. Piotin, and A. F. Frangi: Automated landmarking and geometric characterization of the carotid siphon. *Medical Image Analysis*, vol. 16(4), May 2012, pp. 889-903.

Introduction

- Objective: Describe geometric variability of carotid siphon
 - Large geometric variability prerequisite for being a risk factors
 - 30% of aneurysms occur there
 - Important for stent fitness, endovascular accessibility

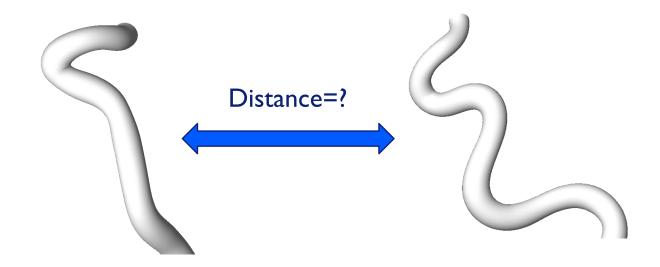




Henry Gray's Anatomy of the Human Body

Introduction

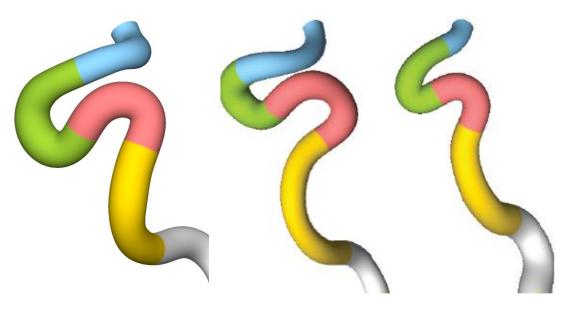
- Carotid siphon shape space
 - Finding a suitable representation and metric
 - Statistical analysis



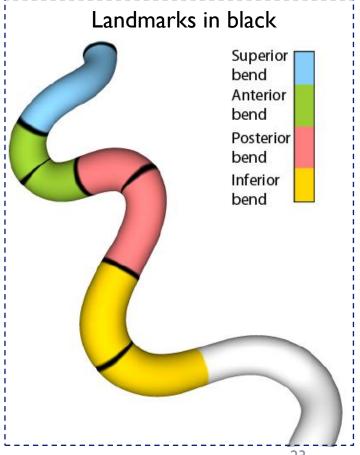
- Option I: Shape described as a vector of geometric features
 - Principal component analysis
- Option 2: Deformation based morphometry
 - Pairwise distance matrix + mapping to Euclidean submanifold

Landmarking

Four bend model



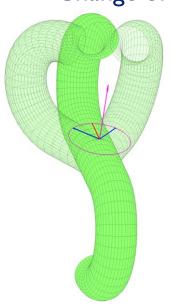
- Landmarks as points of correspondence
 - Define region of interest

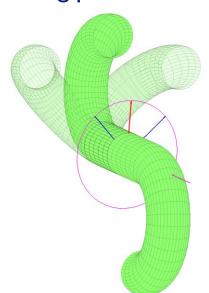


Geometric quantification

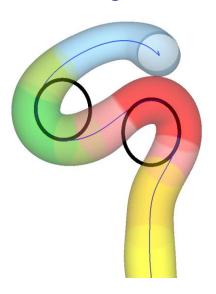
■ Geometric feature vector (scale invariant)

Change of osculating planes



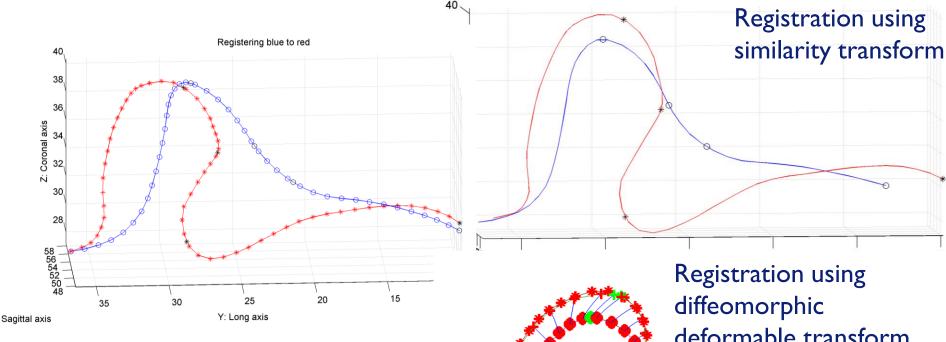






- + global features
 - Tortuosity
 - Bending and twisting energy
 - Curvature and torsion ratio

Deformation based morphometry - LDDMCM



$$J_{C,S}(\phi_t) = \gamma \rho(\phi_t)^2 + E(\phi_1(C), S)$$

$$J_{C,S,\vec{x},\vec{y}}(\phi_t) = \gamma \rho(\phi_t)^2 + \gamma_{cr} E_{cr}(\phi_1(C), S) + \gamma_{lm} E_{lm}(\phi_1(\vec{x}), \vec{y}).$$

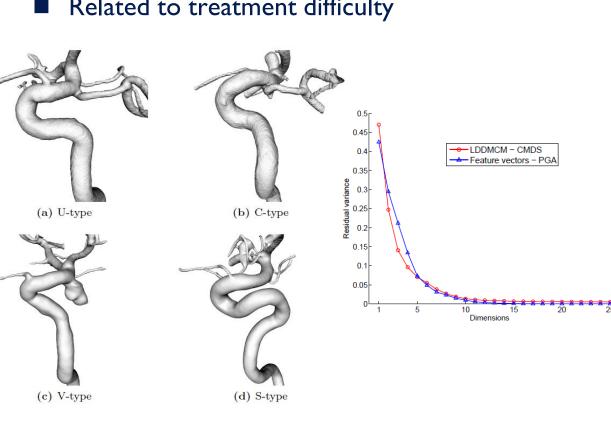
Registration using diffeomorphic deformable transform Glaunes et al. IJCV, 2008

Distance = $\rho(\phi_t)$

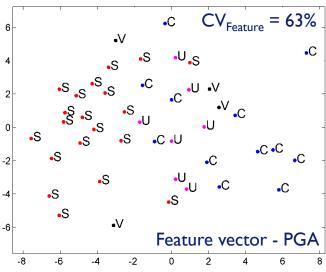
Statistical Analysis

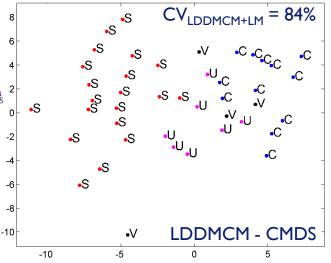
Shape prototypes

- Characterization evaluation
- Related to treatment difficulty



Embedding in 2D Euclidean space





CONCLUSIONS

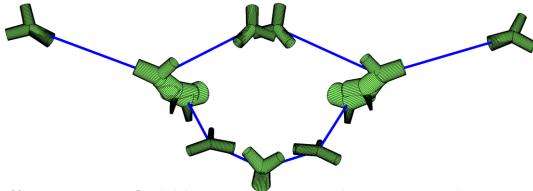
Still open problems:

- Are two vasculatures effectively the same or are they different? (pairwise statistics)
 - Finding a metric in the space of vasculatures (CoWs).
 - Measures that are closely clustered for similar vasculatures and are markedly different otherwise.
- How to model graph/tree-like structures and their variations? (groupwise statistics)
 - Atlas: mean and principal modes of variation (PCA)
 - Change of topology: Data space is strongly non-Euclidean



Start answering questions:

How does the average CoW looks like and what are its principal variations?



- Is there any difference in CoW geometry and topology between men and women?
 - Prevalence of aneurysms is larger in women, could it be due to different CoW geometry between them?
- Is the difference in handedness reflected in the CoW?
- Does the CoW change with aging and how?
- What would be considered normal variability of the CoW and what would be deviations associated with the risk of developing a pathology?
- Can we predict or estimate a risk of aneurysm rupture, from the CoW?

Acknowledgments

Sources of funding



Agència de Gestió d'Ajuts Universitaris i de Recerca



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MINISTERIO DE CIENCIA E INNOVACIÓN



sense and simplicity

Acknowledgments

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CISTIB Center for Computational Imaging & Simulation Technologies in Biomedicine



