# Análisis automático de estructuras tubulares

### Aplicaciones en enfermedades de aorta

*Tahoces PG, Álvarez L, Trujillo A, Cuenca C, González E, Esclarín J, Gómez L, Mazorra L, Alemán-Flores M, Carreira J* 





### Introduction

MDCT  $\rightarrow$  Provides isotropic 3D images, opening the door to the image as a quantitative tool.





## Introduction





## Introduction





### The Aorta Case





### The Aorta Case: Motivation

Segmentation of the aorta is useful for:

- Diagnosis and follow-up, based on several measurements like areas or diameters
- Preoperative planning
- Characterization for classification

Segmentation should be automatic in order to

- Reducing time consuming and efforts made
- Achieve reproducibility





### The Aorta Case







### The Aorta Case





$$
E_1(C, I, \sigma) = \frac{1}{|C|} \oint_C \nabla G_{\sigma} * I(C(s)) \cdot \overline{n}(s) ds
$$



### First contour selection





First contour selection





$$
Q^{z} = \frac{V^{z}}{median\left\{V^{z}\right\}} - \frac{E^{z}}{median\left\{E^{z}\right\}} - \frac{\sigma^{z}}{median\left\{\sigma^{z}\right\}} - \frac{D^{z}}{median\left\{D^{z}\right\}}
$$





**USC** 

$$
\left| E(\alpha,\beta,c) = w_1 E_1(C_{c,\sigma}^{\alpha,\beta}) + w_2 \sqrt{Area(C_{c,\sigma}^{\alpha,\beta})} + w_3 Eccentricity(C_{c,\sigma}^{\alpha,\beta}) \right|
$$

$$
E_1(C, I, \sigma) = \frac{1}{|C|} \oint_C \nabla G_{\sigma} * I(C(s)) \cdot \overline{n}(s) ds
$$
  

$$
C_{c,\sigma}^{\alpha,\beta} = \min E_1(C, I_c^{\alpha,\beta}, \sigma)
$$

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$$

$$
C_{c,\sigma}^{\alpha,\beta} = \min E_1(C, I_c^{\alpha,\beta}, \sigma)
$$
  

$$
I_c^{\alpha,\beta}(x, y) = I \begin{pmatrix} c_x \\ c_y \\ c_z \end{pmatrix} + \begin{pmatrix} \vdots & \cos \alpha & 0 & \sin \alpha \\ -\sin \alpha \sin \beta & \cos \beta & \cos \alpha \sin \beta \\ -\cos \beta \sin \alpha & -\sin \beta & \cos \alpha \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix}
$$

 $u = (\sin \alpha, \cos \alpha \sin \beta, \cos \alpha \cos \beta)$ 





$$
\left| E(\alpha, \beta, c) = w_1 E_1(C_{c,\sigma}^{\alpha,\beta}) + w_2 \sqrt{Area(C_{c,\sigma}^{\alpha,\beta})} + w_3 Eccentricity(C_{c,\sigma}^{\alpha,\beta}) \right|
$$

Descendent Gradient  $f(x)$  $f(X_n)$  $x_{n+1} = x_n - \lambda \cdot f'(x_n)$  $f(X_n)$  $f(X_{n+1})$ Newton-Raphson + Damping Parameter $f(x_{n+2})$  $\sqrt{X_{n+1}}$  $\overline{X_n}$  $X_{n+2}$  $f(x_n)$  $(x_n)$ *n*  $x_{n+1} = x_n - \frac{y_n - y_{n+1}}{n}$  $x_{n+1} = x_{n} - \frac{1}{c_1 c_2}$  $1 - \lambda_n$   $\alpha$  $f'(x_n) + \lambda$  $f'(x_n) + \lambda$ <br>Medical Imaging Pablo García Tahoces  $(x_n)+\lambda$ 

*n*



 $\bf{X}$ 









We have applied the former method to 2 DB's:

- $\cdot$  CDB  $\rightarrow$  Normal and adnormal cases with enhanced contrast
	- 116 Cases  $\rightarrow$  41,182 images
- LDIC  $\rightarrow$  Normal cases with/without enhanced contrast.
	- 290 Cases  $\rightarrow$  54,125 images

























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### Conclusions.

We have developed a fast tracking algorithm especially designed for tubular structures.

The data bases used to verify the performance of the algorithm come from different CTs and various acquisition protocols were employed. The algorithm works well for more than 95% of cases.

We are currently analysing how well performs the algorithm compared with human manual tracing. Preliminary results are encouraging.

