Semi-Automatic Basket Catheter Reconstruction from Two X-Ray Views

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Introduction

Multi-electrode Basket Catheter (Figure 1)

- Multi-electrode catheter to record 64 ECG signals
- Used for localizing the position of electrical anomalies during atrial fibrillation treatment [1]

Methods

Basket Catheter Spline Model



SIEMENS

Basket catheter as seen in bi-plane images. The Figure 1: catheter comprises eight splines carrying eight basket

each catheter spline described using a active shape model [2]

 $p_i = (p_{i1}, \dots, p_{ik}, \dots, p_{i10})^T$ $oldsymbol{x}_{ki} = oldsymbol{R}_i^Y oldsymbol{R}_i (oldsymbol{p}_{ik} + oldsymbol{t}_i) \quad \hat{oldsymbol{x}} = oldsymbol{ar{x}} + oldsymbol{V}oldsymbol{b}$

 $p_{k,j} \in \mathbb{R}^3$: coordinate of kth electrode on ith spline $\bar{R}_{i}^{Y'}$, R_{i} and t_{i} : rotation and translation for alignment \bar{x} : mean shape; V: modes of variation; b: weight factors

• Electrode and Spline Detection

- detect electrodes using determinant of Hessian matrix - detect splines using a vesselness filter [3]

- Point Cloud Generation by Triangulation Correspondence: distance to epipolar line $< M_A$ Motion Compensation: $P'_A = P_A T (1|(t_x, t_y, t_z)^T)$
- Model Initialization and Adaption
 - Length Adaption Most probable model

$$\boldsymbol{b}_k = \operatorname*{argmin}_{\boldsymbol{b}'} \|\tilde{\boldsymbol{p}}_1 - \tilde{\boldsymbol{p}}_{10}\| - \|\hat{\boldsymbol{p}}_1(\boldsymbol{b}') - \hat{\boldsymbol{p}}_{10}(\boldsymbol{b}')\| - a_0 \cdot \mathcal{N}(\boldsymbol{b}'; \boldsymbol{0}, \boldsymbol{\Sigma})$$

Multiple length adaption

$$R = \{ \boldsymbol{r} = c_1 \cdot \boldsymbol{e}_1 + c_2 \cdot \boldsymbol{e}_2 + c_3 \cdot \boldsymbol{e}_3 \} c_1, c_2, c_3 \in \{0, 1.0, -1.0, 1.5, -1.5\}$$

electrodes. Each spline has a marker electrode.



Rotation Initialization

$$\mathcal{D}(\mathbf{b}, \boldsymbol{\alpha}) = a_1 \cdot \underbrace{\left(\sum_{i} \min_{k} d\left(\boldsymbol{e}_i^{\mathrm{A}}, \boldsymbol{S}_k^{\mathrm{A}}(\boldsymbol{b}_k, \alpha_k)\right) + \sum_{i} \min_{k} d\left(\boldsymbol{e}_i^{\mathrm{B}}, \boldsymbol{S}_k^{\mathrm{B}}(\boldsymbol{b}_k, \alpha_k)\right)\right)\right)}_{\text{Distance of each detected electrode to projected splines of the model}}$$

$$a_2 \cdot \underbrace{\left(\sum_{i} \min_{j,k} d\left(\boldsymbol{e}_i^{\mathrm{A}}, \boldsymbol{p}_{k,j}^{\mathrm{A}}(\boldsymbol{b}_k, \alpha_k)\right) + \sum_{i} \min_{j,k} d\left(\boldsymbol{e}_i^{\mathrm{B}}, \boldsymbol{p}_{k,j}^{\mathrm{B}}(\boldsymbol{b}_k, \alpha_k)\right)\right)\right)}_{\text{Distance of each detected electrode to projected electrodes of the model}}$$

$$a_3 \cdot \left(\sum_{i} \min_{j,k} d\left(\boldsymbol{e}_i^{\mathrm{A}}, \boldsymbol{p}_{k,j}^{\mathrm{A}}(\boldsymbol{b}_k, \alpha_k)\right) + \sum_{i} \min_{j,k} d\left(\boldsymbol{e}_i^{\mathrm{B}}, \boldsymbol{p}_{k,j}^{\mathrm{B}}(\boldsymbol{b}_k, \alpha_k)\right)\right)\right)$$

Distance of each projected electrode of the model to detected electrodes

k,j

$$\boldsymbol{\alpha} = \operatorname*{argmin}_{\alpha_1,\ldots,\alpha_8} \mathcal{D}(\mathbf{b},\alpha_1,\ldots,\alpha_8)$$

 $\langle k,j \rangle$

Image-based Model Adaption

$$\mathcal{R}(\boldsymbol{b}, \boldsymbol{\alpha}) = a_4 \cdot \underbrace{\sum_{k} \mathcal{N}(\boldsymbol{b}_k; \boldsymbol{0}, \boldsymbol{\Sigma})}_{\text{Model likelihood}} + a_5 \cdot \underbrace{\sum_{k} \boldsymbol{I}_{\text{ds}}^{\text{A}}(\boldsymbol{S}_k^{\text{A}}(\boldsymbol{b}_k, \alpha_k)) + \boldsymbol{I}_{\text{ds}}^{\text{B}}(\boldsymbol{S}_k^{\text{B}}(\boldsymbol{b}_k, \alpha_k))}_{\text{Distance of projected splines to detected 2-D splines}}$$

 $\boldsymbol{b}, \boldsymbol{\alpha} = \operatorname{argmin} \mathcal{D}(\boldsymbol{b}, \boldsymbol{\alpha}) + \mathcal{R}(\boldsymbol{b}, \boldsymbol{\alpha})$

Figure 2: (a) The coordinates of a subset of the annotated electrodes, normalized and aligned with mean shape of the basket catheter spline model (left) and first two modes of variation. (b) X-ray images of basket catheter with different deformation in phantom and overlay of the reconstructed basket catheter. (c) X-ray images of basket catheter in phantom and overlay of the reconstructed basket catheter shown from left to right (increasing signal-to-noise ratio).

Reconstruction Error in mm Single Marker All Markers C-arm CT 2.2 ± 2.2 2.0 ± 2.0 **Single Marker** (*Figure 2c*) Lab Data 2.0 ± 1.6 1.5 ± 1.2 1.4 ± 1.2 1.3 ± 1.1 **Clinical Data** 4.6 ± 3.5

Discussion and Conclusions

Evaluation

• C-arm CT Data Set

 b, α

18 3-D volumes of basket catheters with different deformations

Laboratory Bi-plane X-Ray with Different Dose

Bi-plane basket catheters X-ray images acquired with different dose setting (signal to noise ratios)

• Clinical Data

Two X-ray views of the basket catheter inside the right atrium

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- Result in phantom data is below the clinically relevant threshold of 3mm [4].
- In largely deformed and twisted real data, larger errors may occur.
- The median error in clinical data set was 3.4mm.
- The overlap of multiple catheters in clinical data introduces error.
- A robust catheter electrode detection is needed.

References

[1] Narayan, S.M. et al., JACC 60(7), 628-636, (2012) [2] Cootes, T.F. CVIU 61(1), 38-59, (1995) [3] Frangi, A.F. et al., LNCV 1496, 130-137, (1998) [4] Bourier, F. et al., JCE 25(1):74-83, (2014)