Segmentation of the Vocal Tract in Magnetic Resonance Images using Deformable Models

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Outline

• Introduction
  – Speech production, Vocal tract, Goals

• Data and Methods
  – MRI protocol, Speech corpus, Vocal tract’s shape modelling (PDM - Point Distribution Model – Building process), Vocal tract’s shape segmentation (ASM - Active Shape Model and AAM - Active Appearance Model)

• Results
  – Landmark points, Modes of variation, Segmentation

• Conclusions
Introduction: Speech production

• Speech production involves complex and individual mechanisms
• It includes three main steps
  – Respiration (air supply – Activation)
  – Phonation (sound origin – Vibration)
  – Articulation (sound modulation – Articulation and Resonance)

Ventura et al. (2009), Application of MRI and Biomedical Engineering in Speech Production Study, Computer Methods in Biomechanics and Biomedical Engineering 12(6):671-681
Introduction: Vocal Tract

- Main shape features
  - Non-linear shape ("L-shaped tube")
  - Significant length
- Articulators’ behavior
  - passive
    (i.e. palate, pharynx)
  - active
    (i.e. tongue, lips, velum)

Introduction: Goals

• **Characterize** and **further reconstruct the vocal tract during the articulation** of (25) **European Portuguese sounds** by using techniques of image processing and analysis on magnetic resonance images

• **Automatically segment the vocal tract** in magnetic resonance images
Data and Methods

• Image acquisition
  – Siemens Magneton Symphony 1.5T system
  – Head array coil

• Speech corpus
  – One training subject (male) in supine position, without speech disorders
  – Twenty five European Portuguese sounds

Data and Methods: MRI protocol

- Slices in sagittal orientation
- T1-weighted images using Turbo Spin Echo sequences
- Time acquisition: 10 sec. (compromise between image resolution and time for sustained articulation tolerable by the subject)
- Static study (sustained articulation during acquisition)
- One image was acquired per each sound (25)

Ventura et al. (2009), Application of MRI and Biomedical Engineering in Speech Production Study, Computer Methods in Biomechanics and Biomedical Engineering 12(6):671-681
Data and Methods: Point Distribution Model (Building process)

Training set (21 images)

(N) Objects sampled by a set of \( n \) (25) landmark points

Building the vector of landmark points’ co-ordinates:

\[ x_i = (x_{i1}, \ldots, x_{in}, y_{i1}, \ldots, y_{in})^T \]

with \( i = 1 \ldots N \)

Pontual Distribution Model

\[ x = \bar{x} + Pb \]

with \( P \) the matrix with the first \( t \) eigen-vectors, and \( b \) the vector of weights for each variation mode

Compute the mean shape \(( \bar{x} )\) and the variation modes (by applying a Principal Component Analysis on the landmark points’ co-ordinates)

Rigid registration of all objects in the training set (considering rotation, translation and scale)

Vasconcelos et al. (2008), Methods to Automatically Built Point Distribution Models for Objects like Hand Palms and Faces Represented in Images, Computer Modeling in Engineering & Sciences 36(3):213-241
Data and Methods: Active Shape Model (Segmentation process)

**Shape Model** (Point Distribution Model)

**Intensity profile** of each landmark point

New image

**Active Shape Model**

**Goal:** match the model into a new image (**image segmentation**)

**By considering iteratively** the steps:
- Search for a better position for each landmark point (through Mahalanobis distance)
- Update the model’s parameters in order to fit the new landmark points’ positions

Data and Methods: Active Appearance Model (Seg. process)

**Shape Model**

**Appearance Model**

Combination of a shape model (geometric) with a texture model (i.e. intensity information)

**Texture Model**

**New image**

**Active Appearance Model**

**Goal**: match the model into a new image (image segmentation)

By considering iteratively the steps:
- Compute the residual errors between the new shape and the model built
- Use the model to predict the changes on its current parameters in order to reduce the residual errors

Vasconcelos et al. (2008), Methods to Automatically Built Point Distribution Models for Objects like Hand Palms and Faces Represented in Images, Computer Modeling in Engineering & Sciences 36(3):213-241
Results: Landmark points

• 25 landmark points were used
  – 4 points at the lips
  – 3 points corresponding to the lingual frenulum and tongue’s tip
  – 7 points equally spaced along the surface of the tongue
  – 7 points along the surface of the hard palate, symmetric with tongue points
  – 1 point at the velum
  – 3 points equally spaced at the posterior margin of the oropharynx

Results: Shape variation modes (-2σ, -σ, 0, +σ, +2σ)

Results: Appearance variation modes \((-\sigma, 0, +\sigma)\)

1\textsuperscript{st} mode of variation

2\textsuperscript{nd} mode of variation

3\textsuperscript{rd} mode of variation
Results: Vocal tract’s shape segmentation

- **Active Shape Model**

- **Active Appearance Model**

Segmentation of the Vocal Tract in MRI using Deformable Models
Results: Vocal tract’s shape segmentation

- Errors between automatic/manual segmentations (in pixel)

<table>
<thead>
<tr>
<th></th>
<th>Image 1</th>
<th>Image 2</th>
<th>Image 3</th>
<th>Image 4</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Active Shape Model</strong></td>
<td>9.99 ± 5.76</td>
<td>9.89 ± 4.43</td>
<td>11.54 ± 6.36</td>
<td>14.23 ± 7.66</td>
</tr>
<tr>
<td><strong>Active Appearance Model</strong></td>
<td>4.90 ± 2.42</td>
<td>10.21 ± 5.09</td>
<td>8.98 ± 4.80</td>
<td>9.91 ± 3.95</td>
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</table>

Conclusions

• We applied statistical models in MR images to analyze the vocal tract’s shape in the articulation of some European Portuguese Sounds and used the models built to segment the vocal tract’s shape in new images.

• From the results experimental results, we can conclude that the statistical models built can extract the main characteristics of the movements of vocal tract effectively.

• Additionally, the models built revealed to be efficient to segment the vocal tract’s shape from MR images.

• Therefore, the models built can be accurate and efficient tools to be used towards the automatic study of the vocal tract from MR images in, for example, speech simulation or rehabilitation.
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