ABSTRACT

When using a set of generic head-related transfer functions (HRTFs) for spatial sound rendering, personalisation can be considered to minimise localisation errors. This typically involves tuning the characteristics of the HRTFs or a parametric model according to the listener’s anthropometry. However, measuring anthropometric features directly remains a challenge in practical applications, and the mapping between anthropometric and acoustic features is an open research problem. Here we propose matching a face template to a listener’s head scan or depth image to extract anthropometric information. The deformation of the template is used to personalise the interaural time differences (ITDs) of a generic HRTF set. The proposed method is shown to outperform reference methods when used with high-resolution 3-D scans. Experiments with single-frame depth images indicate that the method is applicable to lower resolution or partial scans which are quicker and easier to obtain than full 3-D scans. These results suggest that the proposed method may be a viable option for ITD personalisation in practical applications.

Index Terms— HRTF, personalization, depth image, ICP

1. INTRODUCTION

Rendering a spatialised sound source requires encoding localisation cues into the sound signals delivered to a listener’s ear entrances to evoke the perception of the sound coming from a certain direction or location in space. A convenient way to describe the acoustic localisation cues is as a head-related transfer function (HRTF). HRTFs contain the filtering sound undergoes as it propagates from the sound source to the ear entrances due to the presence of the listener’s head and torso. As individual listeners differ from one another anthropometrically, their HRTFs are individual too. Using a generic HRTF set for spatial rendering may reduce fidelity [1, 2]. While using individually measured HRTFs may improve the perceptual quality and accuracy of spatially rendered sound, measuring a listener’s HRTFs is a complex and costly procedure and hence typically not a viable approach for practical applications. Therefore, determining methods to personalise a set of generic HRTFs for a specific listener is of ongoing research interest.

Recent work related to HRTF personalisation includes deriving parametric models [3, 4], best-match selection from a database [5, 6], studying the mapping between anthropometric to acoustic features [1], tuning a low-dimensional HRTF representation [7], direct estimation from 3-D head models [8], and anthropometry-based HRTF modelling and synthesis [9, 10, 11, 12, 13].

An important cue encoded in the HRTF is the interaural time difference (ITD), i.e., the difference in the times of arrival of a sound at both ears. The ITD is a function of the sound’s direction of arrival as well as the listener’s anthropometry. The human auditory system relies on the ITD as a major localisation cue [14, 15]. Precise ITD modelling is crucial for accurate spatial rendering [16]. Commonly used geometric ITD models include the Woodworth model [17] and the spherical scatterer model [18]. Both approximate the head as a rigid sphere. These models have been extended to arbitrary ear angles [19] and ellipsoidal head shapes [20].

Previously, estimating ITDs from a 3-D head scan has been shown [21, 22], and the relationship between morphological changes and HRTF characteristics has been studied [23]. While these approaches showed promising results, their requirement of a full 3-D head scan may be difficult to meet in practical applications. Here we propose a method for personalising ITDs based on fitting a face template to a subject’s complete or partial head scan.
Fig. 2. a) High-resolution 3-D scan with 4096 uniformly distributed surface points (·) and 1583 face points (○); b) face template obtained from 262 high-resolution 3-D head scans.

2. PROPOSED METHOD

The proposed method derives individual ITDs by applying a scaling factor, $s$, to an average ITD contour [12] (see Figure 1). The scaling factor is derived from the deformation of a face template matched to a head scan or depth image of the listener’s head.

2.1. Problem formulation

The HRTF $H(\omega)$ for a specific direction of arrival can be described in the frequency domain as

$$H(\omega) = |H(\omega)|e^{-i\varphi(\omega)},$$

where $\omega$ is the angular frequency, and $\varphi$ is the HRTF phase angle. For practical applications, the unwrapped phase angle $\varphi$ is often considered to be linear, i.e., the ITD $\tau$ is assumed to be frequency independent [24, 19]:

$$\tau = t_L - t_R,$$

where the $t_L$ and $t_R$ denote the time-of-arrival at the left and right ear entrance, respectively. While measured ITDs do exhibit frequency dependence [22], experimental findings indicate that it may not be perceptually relevant [25, 26]. The problem of personalising the ITDs of a generic HRTF consists in mapping the listener’s anthropometric features to the slope of the unwrapped phase angles $\varphi_L$ and $\varphi_R$.

2.2. ITD modelling by scaling average ITD contour

Geometric ITD models are typically based on the assumption that a listener’s ITD contour can be described with a simple model, parameterised via a few anthropometric features, e.g., the head width and the position of the ears. Here, we use an average ITD contour $\bar{I}$ derived directly from measured HRTFs [12]. The average ITD contour of 262 subjects, measured at 400 locations [10] and interpolated at 2048 Fliege points distributed uniformly on the sphere [27, 28], is shown in Figure 1. Given the true ITD contour $I$, derived from a subject’s measured HRTF, the scaling factor $s$ is derived as

$$s = \max \{ \sum_{i=0}^{N-1} (sI_i + k) - I_i \},$$

where $N$ is the number of measurement directions and $k$ an optional bias term.

2.3. Obtaining face template

Given a database of high-resolution 3-D head scans, an average face template $S$ is derived. The scans are aligned so that the head is level, pointing forward, with the centre of the interaural axis lying at the origin of the Cartesian coordinate system (see Figure 2a). Using ray tracing, the vertices $T$ of each 3-D head scan are resampled to give $X_p = [x_0 \ x_1 \ \cdots \ x_P]$, the points of intersection between $T$ and the normals of a cloud of points $P$ uniformly distributed on the surface of a unit sphere [27] (see Figure 2a, black dots). A subset of $G$ face points $F = [f_0 \ f_1 \ \cdots \ f_G] \subset X_p$ is selected that satisfies

$$(\forall i,j) \frac{f_i}{||f_i||} \cdot \frac{v_j}{||v_j||} \geq 0.5,$$

where $(\cdot)$ denotes the dot product, and $V = [x_v \ y_v \ z_v]^T$ is a matrix of directions. Here, $V$ is chosen to roughly point to the front and cheeks of a 3-D scan:

$$V = \begin{bmatrix} 1 & \cos(0.5) & \cos(-0.5) \\ 0 & \sin(0.5) & \sin(-0.5) \\ 0 & 0 & 0 \end{bmatrix}.$$

The green circles in Figure 2a mark the selected face points.

The face template $S = [s_0 \ s_1 \ \cdots \ s_L]$ is obtained by averaging the Cartesian coordinates of the face point sets of all $M$ 3-D head scans:

$$s_i = \frac{1}{M} \sum_{m=1}^{M} f_{i,m}.$$  

The result of averaging 262 scans is shown in Figure 2b.

From the template $S$, a subset of “cheek” points $C = [c_0 \ c_1 \ \cdots \ c_c]$ is obtained that satisfy

$$(\forall i) |y_{c,i}| > 0.06,$$

where $c_i = [x_{c,i} \ y_{c,i} \ z_{c,i}]^T$ and $y_{c,i}$ is given in metres. These points, illustrated by blue circles in Figure 2b, are used to estimate the deformation factor of the matched template, as discussed in Section 2.5.
2.4. Deforming template to match depth image

Given a 3-D head scan or a depth image of the listener’s face taken from a frontal view, a set of semi-uniformly distributed face points \( F \) on the scan surface is obtained via the procedure described in Section 2.3. To align the target face \( F \) with the face template \( S \), an iterative closest point (ICP) transform is performed, yielding the aligned target \( \hat{F} \). The face template \( S \) is matched to the face points \( \hat{F} \) using a nonrigid iterative closest point (NR-ICP) algorithm proposed by Amberg et al. [29]. Figure 3 illustrates the process of fitting the face template to a single depth image.

2.5. Estimating ITD scale from template deformation

Given the deformed template \( \hat{S} \), a deformation factor \( d \) is derived by comparing the \( y \)-coordinates of the subset \( C \) before the NR-ICP transformation and \( \hat{C} \) after the transformation:

\[
y_{\Delta,i} = y_{c,i} - \hat{y}_{c,i}.
\]

The deformation constant is given as:

\[
d_{\Delta} = \text{median} \,(y_{\Delta,i}).
\]

This simple measure for the template deformation was chosen for its robustness when applied to noisy depth images. Alternatively, if high-resolution 3-D scans are used that contain both the subject’s frontal and side views, \( y_{w} \) can be calculated as:

\[
y_{w,i} = || C_{L,i} - C_{R,i} ||,
\]

where \( C_{L} \) and \( C_{R} \) are left and right cheek points, respectively (cf. (6)). The deformation constant is given as

\[
d_{w} = \text{median} \,(y_{w,i}),
\]

in analogy to (8). The deformation factor \( d \) is mapped to the ITD scaling factor \( s \) via linear regression:

\[
s = k_0 d + k_1.
\]

3. EXPERIMENTAL EVALUATION

Ground-truth scaling factors for 180 high-resolution 3-D scans and their corresponding measured HRTFs were calculated via (3). For each of the scans, the face template was matched (see Section 2.4) and the deformation factors \( d_{\Delta} \) and \( d_{w} \) were calculated via (8) and (10). Then, (11) was solved as

\[
s = 4.0849 d_{\Delta} + 1.0064
\]

and

\[
s = 3.9343 d_{w} + 0.4218.
\]

Figure 5 shows the deformation factors of 180 scans plotted against their ground-truth scaling factors. For reference, the correlation between the ground-truth scaling factors and the head-widths \( a \) measured manually with a pair of calipers is shown for 136 subjects, yielding

\[
s = 0.0257 a + 0.6054.
\]

Both deformation factors and manually measured head-widths correlate quite well with the ITD scaling factors.

To assess the accuracy of the proposed method, the errors for estimating the ITD scaling factor as well as the raw ITD estimation errors are calculated. Results are shown in Section 3 for using \( s = 1 \) (“1”), the mean scaling factor (“mean”), \( s \) derived from measured head width (“head width”), and the ground-truth scaling factor (“optimal”). As a baseline, the ITD estimation errors of a parametric spherical head model [20] are provided:

\[
\tilde{\tau} = c^{-1} r_{\text{opt}} (\phi + \sin(\phi)),
\]

where \( c = 343 \text{ m}/\text{s} \) is the speed of sound, \( \phi \) is the lateral angle, and \( r_{\text{opt}} \) is the optimal sphere radius in the least-squares sense. The proposed method yields the lowest errors for the

Fig. 3. Fitting the face template to a (Kinect) depth image; a) raw input depth points; b) depth image after ICP alignment with face template; c) face template after NR-ICP deformation.
scaling factor estimation, and comparable results for the ITD RMSE. ITD_{80} is the RMSE for lateral angles greater than 80 degrees. For the proposed method, it is lower than the ITD RMSE averaged over all directions. This suggests that the scale of individual ITDs is captured by the scaled average contour, but that deviations across lateral angles remain. The results show that the proposed method successfully relates the deformation of the face template to an HRTF feature.

To test the robustness of the proposed method, 123 depth images of the same subject were collected using a Kinect depth camera [30]. The images were collected in two different environments, with slightly differing lighting conditions and head orientations and positions (see Figure 4a). Figure 4b shows the estimated ITD scaling factors. While the estimates seem to exhibit a small bias relative to the ground-truth, especially for \( d_w \), the variance is quite low, indicating that the proposed method is relatively robust across test conditions.

For seven of the subjects in the database with known ITD measurements we collected Kinect depth images similar to the one shown in Figure 3a. Figure 4c shows the estimated ITD scaling factors. It can be seen that they correspond quite well to the ground truth.

## 4. SUMMARY AND CONCLUSION

A method for personalising interaural time differences (ITDs) based on aligning and deforming a face template to a subject’s head scan is proposed. The method is evaluated using a database of high-resolution 3-D head scans and measured head-related transfer functions (HRTFs). As a proof of concept, it is applied to incomplete scans obtained from single-frame depth images. Experimental results indicate that the proposed method performs comparably to using manually measured head-width as an anthropometric feature, and appears relatively robust when used with incomplete head scans. We conclude that applying template matching to depth image data may provide an interesting avenue for personalising HRTFs. Future work includes experimenting with and improving the extraction of a deformation metric from the template matching process and applying it to other aspects of HRTF personalisation.
5. REFERENCES


