Novel Amplitude Scaling Method for Bilinear Frequency Warping-based Voice Conversion

Introduction

- Bilinear frequency warping-based Voice Conversion (VC) [1].
- A novel proposed Amplitude Scaling (AS).
- Limitation of state-of-the-art AS.
- Effectiveness of proposed AS.
- VC Challenge database [4].
- Subjective and objective evaluation of VC systems.

BLFW-based VC

• The allpass transform is given by [1]:

$$Q(z) = \frac{z^{-1} - \alpha}{1 - \alpha z^{-1}},$$
(1)

where $|\alpha| < 1$.

• Frequency warping in cepstral domain.

$$y = W_{\alpha} x, \qquad (2)$$

$$W_{\alpha} = \begin{bmatrix} 1 - \alpha^{2} & 2\alpha - 2\alpha^{3} & \dots \\ -\alpha + \alpha^{3} & 1 - 4\alpha^{2} + 3\alpha^{4} & \dots \\ \vdots & \vdots & \ddots \end{bmatrix}, \quad (3)$$

where W_{α} is a warping matrix (without 0^{th} cepstral) coefficient).

• Conversion function for BLFW+AS:

$$y = W_{\alpha(x,\theta)}x + s(x,\theta), \qquad (4)$$

where
$$\alpha(x,\theta)$$
 and $s(x,\theta)$ are given by,
 $\alpha(x,\theta) = \sum_{k=1}^{m} p_k^{(\theta)}(x) \alpha_k, \quad s(x,\theta) = \sum_{k=1}^{m} p_k^{(\theta)}(x) s_k.$
(5)

- Alignment of source and target feature vectors
- GMM trained on source speaker data, i.e., θ .
- The iterative procedure proposed in [1] for calculating set of $\{\alpha_k\}$ (warping factors) for minimizing the following eq. (6):

$$\epsilon^{(\alpha)} = \sum_{n} ||y_n - W_{\alpha(x_n,\theta)} x_n||^2.$$
 (6)

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State-of-the-art AS Technique

• The $\{s_k\}$ that minimizes the error between warped and target vectors which is given by

$$\epsilon^{(b)} = \sum_{n} ||r_n - s(x_n, \theta)||^2, \qquad ($$

where $r_n = y_n - W_\alpha(x_n)$.

• The least square solution of $P \cdot S = R$ is given by $S_{opt} = (P^T P)^{-1} P^T R.$ (8)

where

$$P_{N \times m} = \begin{bmatrix} p_1^{(\theta)}(x_1) & \dots & p_m^{(\theta)}(x_1) \\ \vdots & \ddots & \vdots \\ p_1^{(\theta)}(x_N) & \dots & p_m^{(\theta)}(x_N) \end{bmatrix}, \quad (9)$$

and $S_{m \times 1} = [s_1 \dots s_m]^T$, $R_{N \times 1} = [r_1 \dots r_N]^T$. (10)

Proposed AS Technique

- Perfect match assumption of state-of-the-art AS.
- Induce spurious peaks.
- Perceptual impression of wrong formant locations.
- Use of GMM-based converted spectrum [3].



Figure 1: Converted spectrum using VC methods. • Propose linear transformation at spectrum-level,

$$\hat{y}_{t}(e^{j\omega}) = \frac{(m_{3} - m_{4})}{(m_{1} - m_{2})}(\hat{x}_{t}(e^{j\omega}) - m_{2}) + m_{4}, (11)$$
where $\hat{x}_{t}(e^{j\omega})$ is the warped only spectrum,
 $m_{1} = max(\hat{x}_{t}(e^{j\omega})), \quad m_{2} = min(\hat{x}_{t}(e^{j\omega})),$
 $m_{3} = max(\hat{x}_{tgmm}(e^{j\omega})), \quad m_{4} = min(\hat{x}_{tgmm}(e^{j\omega})),$
(12)

Presented by Nirmesh J. Shah at 42^{nd} IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP), New Orleans, USA, March 05-09, 2017.

• Total 5 source and 5 target speakers' parallel training data.

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Figure 2: XAB test analysis for voice quality with 95 % confidence interval (margin of error: 0.048 for GMM vs. BLFW+PAS and 0.05 for BLFW+AS vs. BLFW+PAS).

Figure 3: XAB test analysis for speaker similarity with 95 % confidence interval (margin of error: 0.05 for the both cases).

Experimental Setups

• Training set 150 utterances and development set

• Total 25 VC systems for each source-target speaker pair using JDGMM-based method [3], BLFW+AS method and proposed method.

Experimental Results

• XAB test from 375 samples (15 subjects: 5 females and 10 males).





Figure 4: MCD analysis for various systems with 95 % confidence interval (margin of error: 0.04 for all the systems).

• The proposed AS is found to have better voice quality compared to traditional BLFW+AS.

• The proposed system is found to perform less successful in terms of SS after conversion.

• Trade-offs between the quality and the SS.

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The authors would like to thank Dept. of Electronics and Information Technology (DeitY), Govt. of India, for sponsored project, "Development of Text-to-Speech (TTS) System in Indian Languages (Phase-II)" and the authorities of DA-IICT, Gandhinagar, India.









Conclusion

Selected References

Acknowledgements