1. Introduction

The fundamental frequency (F₀) contour refers to the time course of the frequency of the vocal fold vibration of speech. We usually use not only words and sentences to convey messages to the listener in daily communication, but also F₀ contours to add extra flavor to speech such as the identity, intention, attitude, and mood of the speaker. It is also important to note that the naturalness of F₀ contours is one of the most significant factors that affect the perceived naturalness of speech as a whole. In fact, synthesized (artificially created) speech with an unnatural F₀ contour often sounds robotic, lifeless, or emotionless. This article introduces a technique that makes it possible to convert the speaking style of an input utterance into different speaking styles by controlling the F₀ contour while retaining its naturalness.

The proposed technique can be used, for example, to convert the intonation of normal speech to a more lively intonation similar to the way broadcasters speak. It would also allow us to modify the intonation or accents of the acted speech by actors or actresses as desired without the need to retake the scene. It can also be incorporated into text-to-speech (TTS) systems to improve the naturalness of computer-generated speech. Furthermore, we can build a self-training system to assist students in improving their presentation and language skills. We are also interested in applying the proposed technique to develop a speaking-aid system that makes it possible to convert electrolaryngeal speech to normal speech, which can be used to assist people with vocal disabilities.

2. Fundamental frequency contour (intonation and accent)

The F₀ contour of speech consists of intonation and accent components. The intonation component corresponds to the relatively slow F₀ variation over the duration of a prosodic unit, and the accent component corresponds to the relatively fast F₀ variation in an...
accented syllable. Both of these components are characterized by a fast rise followed by a slower fall. The former usually contributes to phrasing, while the latter contributes to accentuation during an utterance. In Japanese, for example, changing the positions of accents results in speech with different dialects or meanings. The magnitudes of these components correspond to how much emphasis the speaker intends to place on the associated phrase or accent. Thus, the magnitudes and positions of these components assist the listener in interpreting an utterance and draw attention to specific words. The F0 contour also plays an important role in conveying to the listener various types of non-linguistic information such as the identity, intention, attitude, and mood of the speaker.

3. Generative modeling of F0 contours

3.1 F0 control mechanism

F0 contours are controlled by the thyroid cartilage, which sits in front of the larynx. The assumption that the F0 contour of speech consists of intonation and accent components is justified by the fact that the thyroid cartilage involves two mutually independent types of movement with different muscular reaction times. Specifically, the intonation and accent components respectively correspond to contributions associated with the translation and rotation movements of the thyroid cartilage. In the late 1960s, Fujisaki proposed a well-founded mathematical model that describes an F0 contour as the sum of these two contributions [1, 2] (Fig. 1). This model approximates actual F0 contours of speech fairly well when the model parameters are appropriately chosen, and its validity has been demonstrated for many typologically diverse languages. If we can estimate the movements of the thyroid cartilage automatically from a raw F0 contour, we can simulate or predict the F0 contour that we may observe when the thyroid cartilage moves differently, by simply modifying the values of the motion parameters. Since the movements of the thyroid cartilage are characterized by the levels and timings of the intonation and accent components, one important challenge is to solve the inverse problem of estimating these components directly from speech. However, this problem has proved difficult to solve. This is because it is difficult to determine a unique intonation and accent component pair only from their mixture (namely an F0 contour), in the same way that it is impossible to determine a unique X and Y pair only from X + Y = 10. Several techniques have already been developed but so far with limited success.

3.2 Statistical formulation

There are some clues that can possibly help solve this problem. That is, we can use the fact that the levels and timings of intonation and accent components are statistically biased in normal speech. The author has proposed constructing a stochastic counterpart of the Fujisaki model that makes it possible to use statistical inference techniques to accurately and efficiently estimate the underlying parameters of the Fujisaki model [3, 4]. The problem of estimating the intonation and accent components from a raw F0 contour is somewhat similar to the audio source separation problem. Audio source separation refers to the problem of separating the underlying source signals from mixed signals. Even though it looks as difficult as solving the X + Y = 10 problem, a statistical approach utilizing the statistical properties and the statistical distribution of the waveforms of audio signals has proved effective in solving it. The idea for the proposed method was inspired by this idea (Fig. 2).

An example of the estimated parameters related to the intonation and accent components (blue and green lines) along with the estimated F0 contours (red line) plotted on the spectrogram of the input speech is shown in Fig. 3(a). An example of converted speech with magnified accent components is shown in Fig. 3(b), and an example of converted speech with shifted positions of accent components is in Fig. 3(c). It is important to note that in these examples we were able to convert the speaking style of the input speech into different styles (one with prominent accents and the other with a different dialect) while retaining the naturalness as if the same speaker were uttering the same sentence in a different way. This was made possible because the proposed framework allows us to modify F0 contours in such a way as never to violate the physical constraint of the actual F0 control mechanism.

![Fig. 1. F0 control model (Fujisaki model).](image-url)
Fig. 2. Proposed model (a stochastic counterpart of the Fujisaki model, described by a discrete-time stochastic process).

For $k = 1, \ldots, K$:

$S_k \sim \pi_{S_{k-1}}$, $S_k$ (state sequence)

\[
\begin{bmatrix}
\mu_i[k] \\
\mu_a[k]
\end{bmatrix} \sim N \left( \begin{bmatrix}
\mu_{i,0} \\
\mu_{a,0}
\end{bmatrix}, \begin{bmatrix}
\sigma_i^2 & 0 \\
0 & \sigma_a^2
\end{bmatrix} \right)
\]

Intonation

\[
G_i[k]
\]

Accent

\[
G_a[k]
\]

$y[k] = x_i[k] + x_a[k]$

Generative model of $F_0$ contours

\[
y \sim N(\mu, \Sigma)
\]

\[
\mu = G_i \mu_i + G_a \mu_a + \mu_0
\]

\[
\Sigma = \sigma_i^2 G_i G_i^T + \sigma_a^2 G_a G_a^T + \Sigma_0
\]

Fig. 3. (a) Example of the estimated parameters related to intonation and accent components (blue and green lines) along with the estimated $F_0$ contours (red line) plotted on the spectrogram of the input speech; (b) example of converted speech with magnified accent components; (c) example of converted speech with shifted positions of accent components.
4. Future perspective

While linear predictive coding (LPC), proposed in the late 1960s, led to the development of a speech analysis/synthesis system that provides a powerful parameter estimation framework for the vocal tract model, the proposed technique enables a new system that does the same for the $F_0$ control model (i.e., the Fujisaki model). In a way similar to the development of LPC, which has given rise to modern speech analysis/synthesis and become the cornerstone module for today’s mobile and voice-over-IP (Internet protocol) communication, this work can also potentially open the door to a brand new speech analysis/synthesis framework.

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References


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