# Fundamental Frequency Contour Reshaping in HMM-based Speech Synthesis and Realization of Prosodic Focus Using Generation Process Model

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## Abstract

Frame-by-frame representation is not appropriate for prosodic features, which are tightly related to speech units spreading a wide time span, such as words, phrases and so on. This causes an inherit problem in fundamental frequency  $(F_0)$  contour generation by HMM-based speech synthesis. Our formerlydeveloped method, which modify generated  $F_0$  contours in the framework of the generation process model, is improved to allow plural phrase components in a breath group. Since the model can clearly relate its commands with linguistic (and para-/non- linguistic) information, the method further enables flexible controls of prosody through manipulating model commands. Prosodic focus is realized in HMM-based speech synthesis as a supplemental process; viewing the differences of command magnitudes/amplitudes between utterances without and with focus. Validity of the method was confirmed by listening experiments of synthetic speech.

**Index Terms**: fundamental frequency contour, generation process model, HMM-based speech synthesis, prosodic focus

# 1. Introduction

Recently, in speech synthesis community, HMM-based speech synthesis attains researchers' special attentions, since it enables a flexible control in speech styles by adapting phone HMMs to a new style. The method processes segmental and prosodic features of speech together in a frame-by-frame manner, and, therefore, it has an advantage that synchronization of both features is kept automatically [1]. Although utterances conveying various attitudes and emotions are possible with rather high quality by the method, frame-byframe processing of prosodic features, however, includes an inherit problem. It has a merit that fundamental frequency  $(F_0)$  of each frame can be used directly as the training data, but, in turn, it generally produces over-smoothed  $F_0$  contours with occasional  $F_0$  undulations not observable in human speech especially when the training data are limited. Moreover, relation of the generated  $F_0$  contours with linguistic (and para-/non- linguistic) information conveyed by them is unclear, preventing further processing, such as to add additional information to speech, to change speaking styles, etc. Prosodic features cover a wider time span than segmental features, and should be treated differently.

One possible solution to this issue is to use the generation process model (henceforth,  $F_0$  model) developed by Fujisaki and his co-workers [2, 3]. The model represents a sentence  $F_0$ contour as a superposition of accent components on phrase ones; each type of components assumed to be responses to step-wise accent commands and impulse-like phrase commands, respectively. These components/commands are known to have clear correspondences with linguistic and para-/non- linguistic information, which is conveyed by prosody. Thus, using this model, a better control can be realized for  $F_0$  contour generation than the frame-by-frame control. Because of clear relationship between generated  $F_0$  contours and linguistic and para-/non- linguistic information of speech, manipulation of generated  $F_0$  contours is possible, leading to a flexible control of prosodic features.

We already have developed a corpus-based method of synthesizing  $F_0$  contours in the framework of  $F_0$  model and have combined it with HMM-based speech synthesis to realize speech synthesis in reading and dialogue styles with various emotions [4]. As an example for the flexible control, we also have developed a method of focus control [5]. Given a speech synthesis system without specific focus control, it is not efficient to prepare a large speech corpus with focus control and train the system from the beginning. The method realizes prosodic focus as a supplemental process to our corpus-based method of  $F_0$  contour generation; train binary decision trees (BDT's) for differences in phrase command magnitudes and accent command amplitudes between utterances with and without focuses. The command values predicted by the baseline method (for utterances without specific focuses) are modified applying the differences. By concentrating on the predicted differences, a better training for  $F_0$  change due to focal position comes possible only with a limited speech corpus. Moreover, speakers for the training speech need not be the same for those of the baseline.

However, in the method,  $F_0$  contours generated by HMMbased speech synthesis are simply substituted by those generated by the method before the speech synthesis. Although, a better quality is obtained for synthetic speech by the method, the segmental and prosodic features are controlled independently, which may cause speech quality degradation due to mismatches between the two types of features. Furthermore,  $F_0$  model commands need to be extracted for utterances included in the training corpus. Automatic command extraction facilitates the process, but is erroneous without (time-consuming) manual correction. Performance of automatic extraction was improved by applying constraints on the command locations to meet the linguistic information of the utterances [6, 7], but it is not enough for making the process fully automatic.

In order to solve this situation, a method has been developed; to reshape the  $F_0$  contours generated by the HMMbased speech synthesis under the  $F_0$  model framework; modifying the  $F_0$  contours to those generated by the  $F_0$  model [8]. This may reduce mismatches between segmental and prosodic features as compared to separately generating both features, though the maximum likelihood condition is not strictly satisfied. The most significant advantage of the method is that  $F_0$  contours are represented by  $F_0$  model commands, and are further controlled easily to realize "flexible" controls in speech synthesis, viz. adding linguistic and/or para-/non- linguistic information, changing speaking styles, and so on. The above-mentioned method of focus control was applied successfully to the modified  $F_0$  contours. The rest of the paper is constructed as follows; after brief explanations on the  $F_0$  model in section 2 and HMM-based speech synthesis in section 3, the method of  $F_0$  contour reshaping is given in section 4. In section 5, focus control is conducted with listening experiments of synthetic speech. Section 6 concludes the paper.

#### 2. Generation process model

Movements of  $F_0$  along time axis are well represented by the  $F_0$  model, which is a command-response model that describes  $F_0$  contours in logarithmic scale as the superposition of phrase and accent components [3]. The *i*<sup>th</sup> phrase component  $G_{pi}(t)$  is generated by a second-order, critically-damped linear filter in response to an impulse-like phrase command, while the *j*<sup>th</sup> accent component  $G_{aj}(t)$  is generated by another second-order, critically-damped linear filter in response to a stepwise accent command:

$$G_{pi}(t) = \begin{cases} \alpha_i^{2} t \ e^{-\alpha_i t} & t \ge 0\\ 0 & t < 0 \end{cases}$$
(1)

$$G_{aj}(t) = \begin{cases} \min\left[1 - (1 + \beta_j t) e^{-\beta_j t}, \gamma\right] & t \ge 0\\ 0 & t < 0 \end{cases}$$
(2)

Based on the analysis of Japanese utterances, time constants  $\alpha_i$  and  $\beta_j$  are known to be almost fixed to 3.0 s<sup>-1</sup> and 20.0 s<sup>-1</sup>, respectively. The parameter  $\gamma$ , which thresholds accent components, can also be set to a fixed value around 0.9. An  $F_0$  contour is then given by the following equation:

$$\ln F_{0}(t) = \ln F_{b} + \sum_{i=1}^{J} A_{pi} G_{pi}(t - T_{0i})$$
  
+ 
$$\sum_{j=1}^{J} A_{aj} \{ G_{aj}(t - T_{1j}) - G_{aj}(t - T_{2j}) \}$$
(3)

where,  $F_b$  is the bias level, I is the number of phrase components, J is number of accent components,  $A_{pi}$  is the magnitude of the *i*<sup>th</sup> phrase command,  $A_{aj}$  is the amplitude of the *j*<sup>th</sup> accent command,  $T_{0i}$  is the time of the *i*<sup>th</sup> phrase command,  $T_{1j}$  is the onset time of the *j*<sup>th</sup> accent command, and  $T_{2i}$  is the reset time of the *j*<sup>th</sup> accent command.

#### 3. HMM-based speech synthesis

HMM-based speech synthesis was conducted for utterances by a male Japanese narrator (MHT) included in ATR continuous speech corpus. Out of 503 sentence utterances, 450 utterances were selected for the HMM training. The utterances were sampled at 16 [kHz] with 16 [bit] accuracy, and windowed with a 25[ms] Hamming window with a 5 [ms] shift. The feature vector consisted of spectral and  $F_0$ parameter ones. The spectral parameter vector consisted of 25 mel-cepstral coefficients (including the 0th one) and their *delta* and *delta*<sup>2</sup> coefficients, while the  $F_0$  parameter vector consisted of log  $F_0$  and its *delta* and *delta*<sup>2</sup> values. Five-state left-to-right HMMs with three emitting states were used. Each state has single diagonal Gaussian output distribution. Decision-tree-based clustering was conducted.

# 4. $F_0$ contour reshaping under $F_0$ model framework

The method of  $F_0$  contour reshaping first decides initial positions of the  $F_0$  model commands from the linguistic information (boundaries and accent types of accent phrases) of

the sentence for speech synthesis, and then estimates their magnitudes/amplitudes from the  $F_0$  contours generated by the HMM-based speech synthesis. The method is similar to the one to find out the  $F_0$  model parameters for an observed  $F_0$  contour [9], but a better extraction is possible. This is because initial positions for the  $F_0$  model commands are decided from the linguistic information, and HMM-based speech synthesis generates  $F_0$  contours free from pitch extraction errors.

The original reshaping method [10] first applies cubic spline interpolation to  $F_0$  sequence  $\hat{p}_c$  generated by the HMM-based speech synthesis to obtain a smoothed continuous  $F_0$  contour for each breath group (period delimited by long pauses (>300 ms)). Then, accent command amplitudes are calculated by taking derivative of the smoothed  $F_0$  contour. Phrase command magnitudes are calculated from the residual of  $F_0$  contour; smoothed  $F_0$  contour minus accent components. Timing parameters are decided from corresponding syllable initial positions by a simple calculation. These parameters are then optimized by the steepest descent method under the following criterion;

$$\hat{\boldsymbol{p}}_{p} = \arg\min_{\boldsymbol{p}} \left( \boldsymbol{p} - \hat{\boldsymbol{p}}_{c} \right)^{\mathrm{T}} U^{-1} \left( \boldsymbol{p} - \hat{\boldsymbol{p}}_{c} \right)$$
(4)

where p is the  $F_0$  sequence generated by the  $F_0$  model. U is the diagonal matrix of variances, which are obtained through the process of  $F_0$  sequence generation in the HMM-based speech synthesis. To avoid over-adjustment, certain limitations are set to parameters during search (Timing commands  $T_{0i}$ ,  $T_{1j}$  and  $T_{2j}$  are searched in ±200 [ms] range from the initial values, for instance). The numbers of commands I and J are fixed to their initial values. Other parameters  $a_i$ ,  $\beta_j$  and  $\gamma$  are fixed to constant values as explained in section 2. All the processes are conducted separately for each breath group.

Finally,  $F_0$ 's are calculated from the  $F_0$  model using the optimized parameters, and are used for the speech synthesis.

Since estimation of phrase commands from the  $F_0$  contour residual is erroneous, phrase commands are assumed only at breath group boundaries with no additional phrase commands in breath groups. This assumption may over-estimate amplitudes of accent commands locating other than breath group initial, but the  $F_0$  contour approximation by the  $F_0$ model is still good and the method improves synthetic speech quality. However, when realizing focus control as a supplemental process, the above situation need to be avoided for a better performance. This is because BDT's for estimating command magnitude/amplitude differences is trained using natural utterances, where phrase commands may occur frequently at breath group medial positions; causing mis-matching between training and application.

In order to solve this situation, search process of  $F_0$  model commands is modified as follows to allow breath-group medial phase commands:

- For each vowel segment, when its  $F_0$  contour has large discontinuities, delete  $F_0$ 's far from  $F_0$  median so that the  $F_0$  variance of the segment comes below a preset value.
- Label vowel segments either L (low) or H (high) according to the accent types of accent phrases. Decide phrase command magnitude for each accent phrase using  $F_0$  values of L segments. The method requires L segments before and after H segment(s). Since no L segments exist before H segment for type 1 accent and after H segment(s) for type 0 accent, they are estimated by subtracting a fixed value from the  $F_0$  value of an H segment (first H segment for type 1

accent, and last H segment for type 0 accent).

- Accent command amplitudes are calculated from the  $F_0$  (residual) value of H segment with the highest  $F_0$ .

The major changes are; no interpolation of  $F_0$  sequences and estimating phrase commands first. The process of parameter optimization is also by the steepest descent method, but counts only vowel segments. This is because sharp undulations out of  $F_0$  model are mostly observed at voiced consonants. The  $F_0$  variances of equation (4) are not counted, since their effect is minor at vowel segments. When phrase/accent command magnitudes/amplitudes come to zero, such commands are deleted. Figure 1 shows the result of  $F_0$  model reshaping by the original and the new methods. The new method generates an  $F_0$  contour with clear declination and with two additional phrase commands at the second breath-group. Informal listening by authors indicates an advantage of the new method.

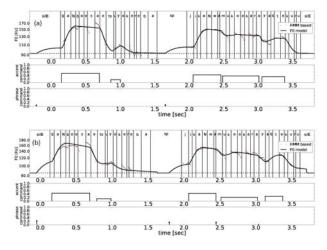


Figure 1:  $F_0$  contour reshaping by the  $F_0$  model approximation for Japanese sentence "gaNboosae tsuyokereba jikaNwa mukookara yattekuru (Time will automatically come if (you) have a strong wish.)."  $F_0$  contours and their  $F_0$  model commands for (a) original method, and (b) new method.

#### 5. Focus control

As mentioned already, one of the most significant advantages of the method of  $F_0$  reshaping is that the modified  $F_0$  contours are represented by the  $F_0$  model, allowing further modifications by changing  $F_0$  model commands. Here, we try to add an emphasis on selected accent phrase to synthetic speech generated by the HMM-based speech synthesis.

Emphasis associated with narrow focus in speech can be achieved by contrasting the  $F_0$ 's of the accent phrase to be focused from those of neighboring accent phrases. This contrast can be achieved by placing a phrase command (or increasing phrase command magnitude, when a command already exists) at the beginning of the accent phrase, by increasing the accent command amplitudes of the accent phrase, and by decreasing the accent command amplitudes of the neighboring accent phrases.

A corpus-based method was developed to predict differences in  $F_0$  model commands between two versions of utterances of the same linguistic content [5, 11]. Applying the predicted differences to the baseline version of speech, the new version of speech can be realized. A large scale speech corpus is not necessary to train  $F_0$  model command differences. This method is applied to realize prosodic focus.

BDT's are first trained predict command to magnitude/amplitude differences between utterances without and with focuses. Since a new phrase command usually appears when a focus is placed where no phrase command initially exists (for an utterance without focus), a phrase command with zero magnitude is assumed there. Tables 1 and 2 show input parameters for BDT's. Fifty sentences were selected from ATR continuous speech corpus, and a female speaker (different from speakers of ATR corpus) was asked to utter them without focus and with focus on one of accent phrases (mostly those including a noun). BDT's are trained using these utterances; 50 utterances without focus and 183 utterances with focus.

Table 1. Input parameters j	for the	prediction	of	differences	in
phrase command magnitude	es.				

Input parameter	Category
Distance of current accent phrase from accent phrase with focus (in number of accent phrases)	5
Number of morae of current accent phrase	3
Number of morae of preceding accent phrase	4
Number of morae of next accent phrase	4
Accent type (location of accent nucleus) of current accent phrase	4
Accent type (location of accent nucleus) of preceding accent phrase	5
Accent type (location of accent nucleus) of next accent phrase	5
Pause immediately before current accent phrase	2 (yes or no)
Magnitude of current phrase command Magnitude of preceding phrase command	Continuous Continuous

Table 2. Input parameters for the prediction of differences in accent command amplitudes.

Input parameter	Category
Distance of current accent phrase from accent phrase with focus (in number of accent phrases)	5
Number of <i>morae</i> of current accent phrase	3
Number of morae of preceding accent phrase	4
Number of morae of next accent phrase	4
Accent type (location of accent nucleus) of current accent phrase	4
Accent type (location of accent nucleus) of preceding accent phrase	5
Accent type (location of accent nucleus) of next accent phrase	5
Pause immediately before current accent phrase	2 (yes or no)
Amplitude of current accent command	Continuous
Amplitude of preceding accent command	Continuous

Figure 2 shows examples of  $F_0$  contours before and after applying the predicted differences to  $F_0$  model command magnitudes/amplitudes. Although prosodic focus involves changes also in pauses and phone durations, they are not counted in the current experiment to see the effect of  $F_0$ contours. Three controls, viz., addition of new phrase command, increase of accent command amplitude for the focused accent phrase, and decrease of accent command amplitudes of neighboring accent phrases, can be seen in the figure.

In order to check the effect of the focus control for realizing emphasis, a perceptual experiment was conducted for the synthetic speech. Eighteen sentences not included in the 50 sentences for training command magnitude/amplitude differences are selected from the 503 sentences of the ATR continuous speech corpus, and two versions of synthetic speech are played sequentially for each sentence. The first one is always utterances without focus (before  $F_0$  change), and the second one is with focus for 16 sentences and without focus for the rest 2 sentences. (These numbers are not known by listeners.) Eleven native speakers of Japanese were asked to listen to these utterances and check an accent phrase where they perceived an emphasis. "No emphasis" answer was allowed. On average, in 77.3 % cases, the accent phrases focused by the proposed method were perceived as "with emphasis."

Modification of  $F_0$  contours may cause degradation in synthetic speech quality. In order to check this point, the same 11 speakers were also asked to evaluate the synthetic speech from naturalness in prosody in 5-point scoring (5: very natural, 1: very unnatural). No apparent degradation is observed from the result; 3.07 (standard deviation 0.96) for utterances with focus and 3.25 (standard deviation 0.93) for those without.

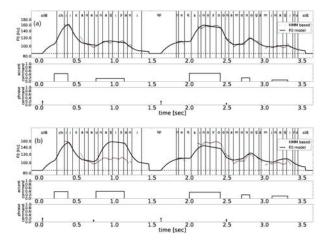


Figure 2:  $F_0$  contours and  $F_0$  model parameters for Japanese sentence "chiisana unagiyani nekkino yoonamonoga minagiru (A small eel shop is filled with a kind of hot air.)." (a) without specific focus and (b) focus on "unagiyani."  $F_0$ contour by HMM-based speech synthesis (without specific focus) is shown for comparison.

## 6. Conclusions

 $F_0$  model is ideal to view  $F_0$  contours in wider time spans with good and clear relations to linguistic and para-/non- linguistic information of speech. A method was developed to increase the naturalness of prosody in synthetic speech generated by HMM-based speech synthesis. It is to modify  $F_0$  contours through representing them with the  $F_0$  model. One of major merits of representing  $F_0$  contours by the  $F_0$  model is that manipulation of prosody is possible as a supplemental process. The method of representing  $F_0$  changes between two styles of utterances as command magnitude/amplitude differences was successfully applied to add prosodic focuses to HMM-based synthetic speech. Better results will be obtainable by inserting pauses (increasing pause lengths when pauses already existing) and changing phone durations. We can also concentrate on differences when handling these [12].

 $F_0$  model stylization may include some problems. One is how to extract  $F_0$  model command parameters automatically for given  $F_0$  contours. The other is that  $F_0$  features are not fully modeled by the  $F_0$  model;  $F_0$  movements due to microprosody are not modeled for instance. This situation is partly the reason of the first problem. These problems can be tackled by two ways; (1) to delete/discard  $F_0$  movements not modeled by the  $F_0$  model when searching  $F_0$  model command parameters, and (2) to keep such  $F_0$  movements as residuals of the  $F_0$  model estimation and treat separately. Also, we plan to apply our method of  $F_0$  contour modification to realize style/voice conversions in HMM-based synthesis.

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