

Time to Frequency Domain Mapping of the Voice Source: the Influence of Open Quotient and Glottal Skew on the Low End of the Source Spectrum

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Abstract

This paper explores the mapping of time and frequency domain aspects of the voice source, focussing on the low end of the source spectrum. It refines and extends an earlier study, where the LF model was used to explore the correspondences between the open quotient (O_q) , glottal skew (R_k) and harmonic levels of the source spectrum, including the H1-H2 measure, widely assumed to reflect differences in O_q . Here we use a different model (the F-model) as it better reflects the effective open quotient and glottal skew in certain conditions. As in the earlier study, a series of glottal pulses were generated, keeping peak glottal flow constant, while systematically varying O_q and R_k . Results suggest that the effects of R_k on the low harmonics is considerably less than estimated in the earlier study, and its main impact is on the level of H2 (and consequently H1-H2) when O_q is relatively high. The conclusion remains that the H1-H2 is not simply a direct reflection of O_q . However, for O_q values of up to about 0.6, it maps closely to H1-H2: beyond this point, H1-H2 reflects a more complex interaction of open quotient and glottal skew.

Index Terms: voice source, open quotient, glottal skew, *H*1-*H*2, frequency domain, time domain, F-model, LF model

1. Introduction

The dynamic modulation of the voice source is a fundamental aspect of speech communication which shapes two essential dimensions of prosody. Firstly, it is an integral part of the linguistic prosody, such as the variations in accentuation, prominence, declination and phrasing that cue information structure. This allows the listener to segment the stream of speech for words and phrases, to identify the important items to attend to in an utterance, etc. The role of voice source modulation in these aspects of linguistic prosody is explored in [1-4]. Secondly, voice modulations carry the paralinguistic prosody, which signals interpersonal information concerning the speaker's state (mood and emotion), attitude to the interlocutor and to the discourse context. These aspects are explored in [5-8].

Despite its communicative importance, this aspect of speech is poorly understood – largely due to the difficulties in obtaining reliable measures of the voice source. The gaps in our understanding have many implications for speech technology. For example, having developed synthetic voices for dialects of Irish [9, 10], the plan to deploy them in interactive educational games and dialogue systems [11, 12] will require being able to approximate the essential linguistic and paralinguistic aspects of voice prosody.

Parameters used in voice analysis are defined either in the time domain or in the frequency domain. Each yields particular insights: time domain measures relate closely to production aspects of the voice, while frequency domain measures relates more closely to perception [13]. Time domain parameters are difficult to estimate reliably, and being phase sensitive, require stringent recording conditions. (For a discussion of analysis difficulties, see [14]). Factors such as these mean that more easily obtained spectral measures are often used as proxy measures of the voice source.

The H1-H2 spectral measure is widely used, often assumed to be an indicator of the pulse open quotient (O_q) [15], and taken as a measure of breathiness [16-21]. An increase in O_q generally leads to an increase in the amplitude of H1, thus increasing H1-H2 [22]. (Note that when the measure is based on the speech waveform, the $H1^*$ - $H2^*$ measure is often used to correct for the vocal tract resonances [23, 15].)

Ideally, one would wish to be able to map reliably between the time and frequency domains to characterise the source (see, for example [21, 13, 24, 25, 26]). In this paper, we build on an earlier study [26], which focussed on the low end of the source spectrum, and which used LF model [27] simulations to explore the correlation of time domain parameters with the levels of the first three harmonics. The time domain parameters considered included the peak flow (U_p) – described by Fant and Lin [24] as the principle determinant of H1 – and the pulse shape parameters open quotient (O_q) and glottal pulse skew (R_k) . O_q is the duration of the open phase as a proportion of the glottal period $(O_q = (t_p - t_o)/T_0)$ and R_k is a measure of glottal pulse symmetry, given by the duration of the closing branch of the pulse relative to the duration of the opening branch. Thus, a lower R_k value corresponds to a more skewed pulse $(R_k = (t_e - t_p)/(t_p - t_o)$, see Figure 1).

By maintaining a constant U_p and by varying O_q and R_k systematically, the aim was to illuminate how these parameters affect H1 and H2, as well as the H1-H2 measure. Although O_q emerged as the main determinant of the amplitudes of H1 and H2, there was a strong influence of glottal skew, particularly on H2. With regard to H1-H2, although a broad correspondence with O_q , emerged, glottal skew was found to have a strong effect, particularly at high O_q values. It is clear that these effects can potentially invalidate direct inferences from H1-H2 on O_q and on voice quality.

However, a re-examination of the LF model simulations in [26] lead us to reconsider these findings, and motivate the present study. As illustrated in Figure 1, an LF pulse with very high O_q (0.85) and high skew (low R_k of 0.15) yields an effective O_q that is considerably lower than the numerical value of the model (0.5 rather than 0.85). This is due to the fact that during the initial part of the opening phase of the LF pulse, the flow is very close to zero – a consequence of the exponentially

growing sinusoidal segment used to model the open phase in the LF model (for details, see further [26-29]).

In Figure 1, an estimate of the *effective* open quotient (OQ_e) and skew (RK_e) for these LF settings is obtained by fitting a triangular pulse to the LF glottal pulse. This triangular pulse is made up of two line segments. The line for the opening branch intersects with the model at the point of the maximum slope and at the point of peak glottal flow. Similarly, the line for the closing branch intersects with the point of maximum slope and the peak glottal flow. There are of course many possible stylisations, but this simple stylisation is deemed adequate to capture the effective open quotient.

As already pointed out, Figure 1 (left panel) highlights how much smaller the effective open quotient ($OQ_e = 0.50$) of the LF pulse is than its numeric value ($O_q = 0.85$) would imply. Differences also emerged in the glottal skew, where the effective R_k values are higher than the specified values in the model.

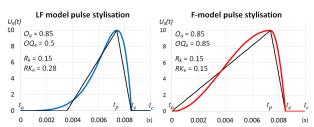


Figure 1: LF model pulse (left panel) and F-model pulse (right panel) with triangular stylisations.

For the reanalysis in this paper, we used the three-parameter model described in [30, 31] (see also the specification of the model in Section 2 below). In [27] this model is referred to as the *F-model*, a term which is adopted henceforth.

A similar stylisation of the glottal pulse for the F-model is also shown in Figure 1 (right panel) using the same settings for O_q and R_k as were used for the LF pulse (left panel). As can be observed there, the F-model generates pulses (red) where O_q and R_k values are always the same as the effective open quotient and glottal skew values of the stylised pulse (black). It is striking, for the same O_q and R_k settings, how divergent the effective open quotients are in the two models in this case.

2. Methodology

In [26] we used the LF voice source model to analyse the relationships between the glottal pulse shape and the amplitudes of source spectral components. However, for reasons outlined above, we here carry out essentially the same analysis, but instead using a different source model.

Thus, we used the F-model to generate a range of glottal pulses for which the open quotient and glottal pulse skew were varied in controlled steps by changing the O_q and R_k parameters, while the peak amplitude ($U_p = 10$) and pulse duration were kept constant.

We also reanalysed the data from our earlier study [26] using OQ_e and RK_e for the LF model (instead of O_q and R_k) for the correlation with H1 and H2 amplitudes.

The F-model (see Figure 1, right panel) is determined by the two expressions in (1), which generate the glottal pulse of the opening phase and closing phase respectively. The model does not include a return phase after the main excitation.

$$U_{g}(t) = \begin{cases} \frac{U_{p}}{2} (1 - \cos \omega_{g} t) & t_{o} \leq t \leq t_{p} \\ U_{p} (1 - K(1 + \cos \omega_{g} t)) & t_{p} < t \leq t_{e} \\ 0 & t_{e} < t < t_{c} \end{cases}$$
(1)

In addition to f_0 , the F-model has three parameters: U_P , ω_g and K. $\omega_g = 2\pi F_g$ where F_g is the characteristic frequency of the pulse. K determines the rate by which the flow drops during the closing branch of the pulse. Neither O_q nor R_k are parameters of the F-model, but ω_g and K can be derived from O_q and R_k using the following formulas: $\omega_g = \pi f_0 O_q^{-1} (1 + R_k)$ and $K = (1 - \cos \pi R_k)^{-1}$.

As in [26], glottal pulses were generated with nine different O_q settings, ranging from 0.15 to 0.95 in steps of 0.1. For each O_q setting, nine pulses with different R_k values were generated. The R_k values also ranged from 0.15 (high skew) to 0.95 (low skew) in steps of 0.1, thus covering most of the possible R_k range.

The sampling frequency used was 20 kHz, which ensured that the effect of aliasing on the lower end of the source spectrum would be negligible. Each pulse was repeated five times in order to produce a harmonic spectrum. For each of the 81 glottal waveforms, a 1000-point (50 ms, rectangular window) DFT spectrum was calculated and the amplitudes of the first two harmonics were extracted. The window length was chosen so that the output frequency samples would coincide with the harmonic frequencies, thus avoiding potential rounding errors.

3. Results

The left panels of Figure 2 show the H1 and H2 amplitudes as a function of O_q and R_k , when derived from the F-model. The previously reported estimates [26] using the LF model are shown in the mid panels. Shown in the right panels are the values obtained for the LF model when OQ_e and RK_e are used instead of O_q and R_k for the mapping to H1 and H2. OQ_e and RK_e are the effective values of the open quotient and glottal skew respectively, according to the stylisation in Figure 1.

It is clear that the effects of R_k on the H1 and H2 amplitudes are reduced when the source spectrum is derived from the F-model rather than from the LF model. Note for H1 that the R_k -lines are much closer together, and while there is still some R_k influence, the differences are very small compared to the differences found for the LF model spectra, particularly when R_k is between 0.15 and 0.35. Note also that in the F-model spectra the peak in H1 occurs close to $O_q = 0.79$ regardless of R_k , quite unlike in the LF model spectra where the peak occurs at increasing O_q values as R_k decreases.

Similarly for H2, the effects of R_k are much smaller in the F-model spectra compared to the LF model spectra. The peak H2 values for the F-model spectra occur consistently around $O_q = 0.37$, which is strikingly different to what was previously found in the LF model spectra (lower mid panel) where the peak in H2 occurs at increasing O_q values when R_k is reduced.

All in all, what emerges is a simpler and more predictable correlation of the time and frequency dimensions. Nonetheless, there are still some clear differences in H2 levels, depending on the degree of glottal skew, particularly for higher O_q values.

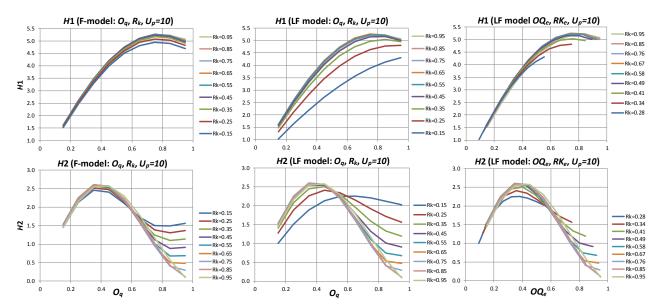


Figure 2: Variation in H1 and H2 as a function of O_q and R_k for the F-model (left panels) and for the LF model (mid panels). The right panels show H1 and H2 variation as a function of the effective open quotient and glottal skew measures OQ_e and RK_e (see Figure 1) for the LF model pulses. $U_p = 10$ in all cases.

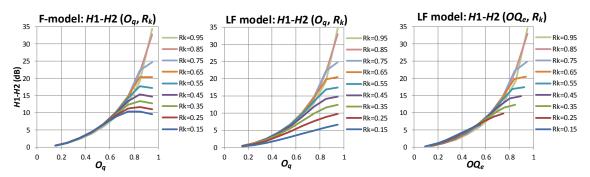


Figure 3: Variation in H1-H2 as a function of O_q and R_k for the F-model (left panel) and for the LF model (mid panel). The right panel shows H1-H2 variation as a function of the effective open quotient and glottal skew measures OQ_e and RK_e for the LF model pulses.

In the right panels of Figure 2, the replotted LF model data for H1 and H2 are shown. Using the effective values of open quotient and glottal skew $(OQ_e \text{ and } RK_e)$ the H1 and H2 variation that emerges is much more like that found for the F-model.

3.1. H1-H2 as a measure of the open quotient

Figure 3 illustrates the H1-H2 variation as a function of O_q and R_k derived from the F-model spectra (left panel) and from the LF model spectra (mid panel). The right panel of Figure 3 shows the H1-H2 variation as a function of OQ_e and RK_e derived from the LF model. In the case of the spectra derived from the LF model, the H1-H2 measure is highly correlated with O_q , but there is also a strong interaction with R_k , as shown previously also in [26, 32, 33]. In comparison, for the F-model spectra, the H1-H2 value is essentially determined by O_q up to about $O_q = 0.6$, with little influence of R_k . For higher O_q values one again notes the strong interaction with R_k . What this effectively means is that if the H1-H2 value is below 8 dB, it can be reasonably assumed to reflect O_q differences. For level differences above 8 dB these simulations imply a combined influence of O_q and R_k : unless one is given, the

other cannot be inferred. As in Figure 2, we note that when replacing original O_q and R_k values with the effective values OQ_e and RK_e , the LF model data mirror the F-model data more closely.

4. Estimating the correction factor, k

As discussed in Fant and Lin [24], the amplitude of H1 is proportional to U_p and can be derived according to (2), where |R(f)| represents the radiation function and k is a correction factor which depends on the glottal pulse shape.

$$H1 = k \cdot \frac{U_p}{2} |R(f)| \tag{2}$$

In [26] we used spectral measurements from the LF model simulations to determine the k factor. It was found that the variation in H1 due to changes in O_q closely follows part of a parabolic curve. Thus, a quadratic function was presented for the prediction of the k factor according to:

$$k = a_2 O_q^2 + a_1 O_q + a_0 (3)$$

The coefficients of the polynomial in (3) depend on R_k , and the variation in the three *a*-coefficients were shown to also closely match quadratic functions. Thus, the *a*-coefficients can be determined by the expressions in (4).

$$a_{2} = b_{2}R_{k}^{2} + b_{1}R_{k} + b_{0}$$

$$a_{1} = c_{2}R_{k}^{2} + c_{1}R_{k} + c_{0}$$

$$a_{0} = d_{2}R_{k}^{2} + d_{1}R_{k} + d_{0}$$

$$(4)$$

If we now use instead the spectral measurements from the F-model, we find that the expressions in (3) and (4) are still valid ($R^2 = 0.97$ or higher). The values of the nine coefficients in (4), which are used to calculate the *a*-coefficients, are shown in Table 1

Table 1: Values of the coefficients for the equations in (4) derived from the F-model voice source data.

b_2	b_1	b_0
1.04	-1.03	-1.58
c_2	c_1	c_0
-1.17	1.27	2.52
d_2	d_1	d_0
0.0505	-0.0619	-0.0567

Note, however, that the R_k -dependent variation in the a-coefficients is quite different when the data are based on the F-model: the variation is much smaller, particularly for a_1 and a_2 (see Figure 4). It is therefore feasible to use constant a-coefficients, independent of R_k . By calculating the mean values over the full R_k range we get the constants shown in (5), which can be used without losing too much accuracy in the calculation of the k factor.

$$a_2 = -1.8$$

 $a_1 = 2.8$ (5)
 $a_0 = -0.07$

To test the accuracy of the of k correction factor as determined by (3) and (4), the H1 prediction errors were calculated on a different set of 64 F-model pulses, where both O_q and R_k varied from 0.2 to 0.9 in steps of 0.1. The correlation between the estimated and actual H1 amplitudes is very high ($R^2 = 0.9995$) with mean and maximum errors in the H1 estimate being 0.07 dB and 0.28 dB respectively. Using the constant coefficients in (5), the corresponding error values are 0.19 dB (mean error) and 0.47 dB (max error) and $R^2 = 0.994$.

5. Conclusions

This paper refines and extends an earlier study [26] concerning the mapping of time to frequency domain measures of the voice source. It is found that the influence of the glottal skew on the first two harmonics is not as great as was thought in [26], but was overestimated due to the way the open phase of the LF model is generated, which results in lower *effective* open quotient values in certain conditions. In the present revised estimates, it emerges that H1 can be determined reasonably accurately from U_p and O_q , independently of glottal skew. The appropriate k factor for estimating H1, based on these revised calculations is presented.

H2 is also less dependent on the glottal skew than was thought in [26], but there is still a clear influence, particularly when the open quotient is high.

Variation in R_k dependent

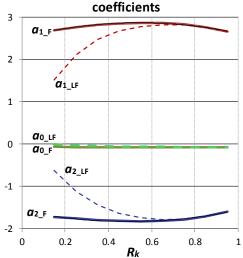


Figure 4: Variation in the R_k dependent a_2 , a_1 and a_0 coefficients of (4) for the F model (solid lines) and the LF model (dashed lines).

With regard to H1-H2, when the difference is below about 8 dB, the correlation with O_q is straightforward, and there is little influence of glottal skew. When the difference exceeds this value, there is a clear interaction with the pulse skew, where increasing skew counteracts the effects of increasing open quotient.

Thus, in broad terms our findings agree with previous studies in that H1-H2 is not uniquely attributable to O_q , but reflects a more complex interaction of open quotient and glottal skew. However, the present results show a more nuanced picture of this interaction, showing that the influence of glottal skew may in fact be negligible when the open quotient is relatively low.

This study draws attention to differences between voice source models in the way that parameters such as O_q and R_k are realised, which can yield differences in their effective values. It may therefore be useful to work with effective measures of open quotient and glottal skew in controlling the source model. For a comparison of different measures of open quotient, see [34].

The focus here is limited to the parameters that shape the low end of the source spectrum. This is thus only a small step towards a broader enterprise, which is the overall mapping of the time-to-frequency correspondences in the glottal source. Future work will address parameters that govern the mid and upper parts of the source spectrum.

To conclude, an eventual mapping between time and frequency dimensions of the source would confer many benefits. It would extend our understanding of the production and perception of voice quality and bring us closer to developing more robust voice source analysis techniques, crucial to the goal of modelling voice prosody in human speech and replicating it in technology.

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7. References

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