



Computation of L2 speech rhythm based on duration and fundamental frequency

Pellegrino, Elisa; He, Lei; Dellwo, Volker

Abstract: Rhythmic characteristics of speech vary between native and non-native speakers. Studies comparing the rhythmic properties of L1 and L2 speech based on rhythm metrics have shown that this relationship is far from straightforward. It seems evidently the case that the difference between native and non-native speech is a complex interaction of a variety of rhythmic cues (duration, F0 and intensity). In this study we extended the durational domain by F0 and tested whether metrics combining duration and F0 (henceforth combined measures) could better account for the rhythmic differences between L1 and L2 speech. To test this, 5 native Mandarin speakers and 5 Italian learners of Mandarin recorded *The North Wind and the Sun* in Mandarin. Besides, each Italian speaker also recorded the same text in Italian. Each sentence in L1-L2 Chinese and in L1 Italian was segmented into syllables. We calculated duration- and F0-based metrics (Δ syllable duration/F0; r-PVI syllable duration/F0; Varco syllable duration/F0; nPVI syllable duration/F0), as well as combined metrics (Δ combined, Varco combined, rPVI combined and nPVI combined). Results show that measures taking syllable duration and F0 separately do not differentiate consistently between L1 and L2 Chinese, while combined metrics reflect more systematically the variability pattern between L1 and L2 Chinese speech rhythm. More research is, however, needed to explore the significance of combined metrics as opposed to duration- and F0-based measures for the perceived rhythmic differences between L1 and L2 speech.

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VORWORT

Der vorliegende Tagungsband dokumentiert die 28. Konferenz „Elektronische Sprachsignalverarbeitung (ESSV)“. Zum ersten Mal findet die ESSV in Saarbrücken auf dem Campus der Universität des Saarlandes statt. Zweck der ESSV ist es, Interessenten aus dem Gebiet der Sprachtechnologie in Forschung und Anwendung zusammenzubringen. Daher eignet sich der Saarbrücker Uni-Campus als Konferenzort in besonderer Weise, da mit der Fachrichtung Sprachwissenschaft und Sprachtechnologie (mit den Schwerpunkten Computerlinguistik und Phonetik) dem DFKI (Deutsches Forschungszentrum für Künstliche Intelligenz), dem Exzellenzcluster „Multimodal Computing and Interaction (MMCI)“ sowie dem Sonderforschungsbereich „Information Density and Linguistic Encoding“ eine hohe Dichte an für die Sprachtechnologie relevanten Institutionen vorhanden ist.

Die hohe Anzahl an Einreichungen, die wir erhalten haben, lässt darauf schließen, dass die breit gefächerte Thematik nach wie vor ein großes Interesse hervorruft – und das nun in ihrem 28. Jahr als deutschlandweite Tagung. Wir nehmen auch mit Freude zur Kenntnis, dass die Anzahl der Beiträge im Vergleich zu den Vorjahren gestiegen ist und dies zum einen auf eine große Vernetzung und zum anderen auf zahlreiche Einreichungen junger Forscherkollegen zurückzuführen ist. Daher sehen wir die ESSV auf einem guten Weg.

Das Rückgrat der Tagung wird durch fünf Hauptvorträge gebildet, die auch Leitbilder der Tagung sein sollen. Yves Laprie (Nancy) spricht zur artikulatorischen Modellierung des gesamten Vokaltraktes basierend auf medizinischen bildgebenden Verfahren. Dazu passt die thematische Sitzung zur physischen Modellierung der Sprachproduktion, die Peter Birkholz dankenswerterweise initiiert hat. Stefan Kleiner (Mannheim) berichtet über groß angelegte Studien zur regionalen Variation in der deutschen Standardaussprache. Regionale Varietäten sind aus sprachtechnologischer Sicht hochinteressant; dem wird in der Sitzung zu Sprachsynthese Rechnung getragen. Sprachmodellierung im Allgemeinen steht im Fokus der Sitzung, zu der Dietrich Klöcker (Saarbrücken) einen Hauptvortrag hält. Bei Dialogsystemen handelt es sich um ein weiteres klassisches Feld der Sprachtechnologie. Jan Alexandersson (Saarbrücken) wird in einem aktuellen Projekt zu multimodalen Aspekten von Dialogmodellierung berichten. Der letzte eingeladene Vortrag wird von Tanja Schultz (Bremen) zum Thema gesprochene Kommunikation basierend auf Bio-Signalen bestritten, welches um weitere reguläre Beiträge bereichert wird. Die Felder Verarbeitung von Affektivität in gesprochener Sprache sowie kognitive Modellierung runden die mündlichen Präsentationen ab. Die breite Themenpalette setzt sich in den mehr als 30 Posterpräsentationen fort.

Wir freuen uns darauf, Sie in Saarbrücken begrüßen zu dürfen, und wir wünschen allen Teilnehmern eine erfolgreiche Konferenz in einer angenehmen und kollegialen Atmosphäre.

Saarbrücken, im Februar 2017

Jürgen Trouvain, Ingmar Steiner, Bernd Möbius



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**AUDIO COMPRESSION AND ITS IMPACT ON EMOTION RECOGNITION IN
AFFECTIVE COMPUTING**

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Abstract: Enabling a natural (human-like) spoken conversation with technical systems requires affective information, contained in spoken language, to be intelligibly transmitted. This study investigates the role of speech and music codecs for affect intelligibility. A decoding and encoding of affective speech was employed from the well-known EMO-DB corpus. Using four state-of-the-art acoustic codecs and different bit-rates, the spectral error and the human affect recognition ability in labeling experiments were investigated and set in relation to results of automatic recognition of base emotions. Through this approach, the general affect intelligibility as well as the emotion specific intelligibility was analyzed. Considering the results of the conducted automatic recognition experiments, the SPEEX codec configuration with a bit-rate of 6.6 kbit/s is recommended to achieve a high compression and overall good UARs for all emotions.

1 Introduction

In affective computing the assumption is made that machines need to be capable of expressing and recognizing affects to enable a natural Human Computer Interaction. In the last few years the focus shifted from “acted” to “in the wild” emotion analyses [1]. This term is often linked to the concept of realistic emotions, but however, mobile applications with all their limitations are still out of focus. Most approaches for an automatic speech-based affect recognition still utilize uncompressed, high-quality speech [2], although speech compression techniques have shown significant impact on acoustic characteristics [3].

Until now, the effects of speech compression on affective speech have been rarely addressed. To do so, one can rely on the well-established base emotions which have obvious ground truth labels, where recognition results on uncompressed data are well known and which therefore can serve as benchline. One of the first studies presented in [4] analyzes the impact of various speech related codecs on emotion recognition performance of Gaussian Mixture Models (GMMs) using only fear-type emotions. The authors of [5] use GMMs as well to analyze the impact of compression on emotion recognition using different acoustic feature types. Unfortunately, both studies focused on the pure recognition results without investigating the underlying spectral error introduced by the compression, although it is known from speech-based emotion recognition studies, that the spectral information is very important to achieve high performance [6]. Additionally, from codec design research, the listening quality assessment is often used to analyze the speech quality [7]. Both evaluations will be investigated within this paper using appropriate measures. Another neglected aspect is the human recognition ability of affective compressed speech. It gives a feeling of the degradation of acoustic characteristics, regardless of the possible technical insufficiencies of the current imperfect automatic emotion recognition. Both aspects, the spectral error and the human affect recognition ability of compressed speech are investigated in the current paper and set in relation to the automatic emotion recognition

COMPUTATION OF L2 SPEECH RHYTHM BASED ON DURATION AND FUNDAMENTAL FREQUENCY

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Abstract: Rhythmic characteristics of speech vary between native and non-native speakers. Studies comparing the rhythmic properties of L1 and L2 speech based on rhythm metrics have shown that this relationship is far from straightforward. It seems evidently the case that the difference between native and non-native speech is a complex interaction of a variety of rhythmic cues (duration, F0 and intensity). In this study we extended the durational domain by F0 and tested whether metrics combining duration and F0 (henceforth combined measures) could better account for the rhythmic differences between L1 and L2 speech. To test this, 5 native Mandarin speakers and 5 Italian learners of Mandarin recorded The North Wind and the Sun in Mandarin. Besides, each Italian speaker also recorded the same text in Italian. Each sentence in L1-L2 Chinese and in L1 Italian was segmented into syllables. We calculated duration- and F0-based metrics (Δ syllable duration/F0; r-PVI syllable duration/F0; Varco syllable duration/F0; nPVI syllable duration/F0), as well as combined metrics (Δ combined, Varco combined, rPVI combined and nPVI combined). Results show that measures taking syllable duration and F0 separately do not differentiate consistently between L1 and L2 Chinese, while combined metrics reflect more systematically the variability pattern between L1 and L2 Chinese speech rhythm. More research is, however, needed to explore the significance of combined metrics as opposed to duration- and F0-based measures for the perceived rhythmic differences between L1 and L2 speech.

1 Introduction

Over the past few years language-specific rhythmic characteristics have been associated with the durational properties of consonantal and vocalic intervals, and a wide array of interval-based measures (henceforth rhythm metrics) have been developed to quantify between-language rhythmic differences. [1] proposed to measure the proportion over which speech is vocalic (%V) together with the variability of consonantal and vocalic intervals over an entire utterance (ΔC , ΔV). [2] suggested to calculate the average durational differences between two consecutive vocalic or consonantal intervals (r-PVIC, n-PVI-V). Some of the original metrics were further developed: [3] and [4], for example, designed the normalized variants of ΔC and ΔV (varcoC and varcoV) to control for speech rate variations. (For a detailed overview of the various metrics, see [5]).

The very nature and robustness of these metrics for the rhythmic classification of languages have been recently disputed for a variety of reasons: a) they focus exclusively on durational properties, thus reducing rhythm to timing; b) they disregard the role played by other prominence-bearing acoustic parameters in the creation of speech rhythm, c) they provide rhythmic values to languages that varied considerably across different data collection methods, sentence materials, segmentation procedures and speaking styles [6]. There have subsequently been developed alternative models of speech rhythm which focused on prominence-lending parameters other than duration, such as intensity [7], [8], [9], fundamental frequency [10], loudness [11], [12], and the frequency components of the amplitude envelope [13].

Which model of speech rhythm has been employed in research on second language speech? Despite the reservations about the rhythmic metrics, the approach that has been most widely applied in second language research is the one focusing on the temporal characteristics of consonantal and vocalic intervals. Although there is evidence that native and non-native speech vary considerably in terms of their segmental durational characteristics (e.g. [14]-[15] and [16]), the search for systematic rhythmic differences between L1 and L2 speech, or among learners at different proficiency levels has provided inconclusive results. The findings across studies seem largely to depend on the rhythm metric and on the L1/L2 pair under investigation.

In a review of the relevant literature, [17] found that %V distinguished native speakers of English from French and Spanish learners, but not from Japanese learners [15], [16], [18]. The consonantal and vocalic PVI displayed different values between Korean learners of English and English native speakers [19], but did not separate either Spanish-accented English from L1 English or English-accented Spanish from L1 Spanish [15], [20]. Similar divergences can be derived from the literature on the development of L2 rhythm. For instance, [21] and [22] found that rate normalized metrics distinguished beginner, intermediate and advanced learners of English with German and French as their L1s. Conversely, [19] did not find any significant differences between Korean learners of English at different proficiency levels.

This considerable disagreement across L2 studies further undermines the validity of rhythm metrics as instruments to quantify the rhythmic properties of languages. Given that L2 speakers can sound rhythmically dissimilar from the native speakers even for deviations in intonation, stress, pitch range and intensity [23], [24], [25], a more suitable approach to compare L1 and L2 speech rhythm or to trace the acquisition of L2 rhythm should integrate the contribution of acoustic correlates of prominence other than duration. In light of previous findings showing that duration and F0 are interdependent cues to perceived rhythm [26], in the present study we extended the durational domain by F0 and tested whether metrics taking into duration and F0 (henceforth combined metrics) capture more reliably the rhythmic variability between L2 and L1 speech than metrics based only on either duration or F0.

2 The Experiment

2.1 Participants, materials and recordings

5 Italian learners of Mandarin and 5 native Mandarin speakers were recorded in the anechoic room of the Language Center of the University of Naples L'Orientale, while performing a reading task. Both groups read *The North Wind and the Sun* in Mandarin. Each Italian native speaker also recorded the same text in Italian. Native Italian speakers were all university students from the Campania region, ranged in age from 21 to 23, had studied Chinese for three years only in a classroom setting. According to the language competence self-assessment grid (CEFR [27]), they were pre-intermediate learners (B1). The native Mandarin speakers ranged in age between 25 and 36. They graduated in China in humanities and were advanced learners of Italian (C1). Given the gender-related differences in f_0 values, both groups of informants were composed of only female speakers.

2.2 Segmentation and measurements

In order to compute duration, F0 and combined metrics (duration and F0), each sentence in L1 - L2 Chinese and in L1 Italian was segmented into syllables using Praat [28]. The syllabification criteria were more acoustic than phonological. Syllabification of Mandarin was based on the sequence of phones that correspond to each orthographic character (fig. 1). However, where two stops met or one stop was left adjacent to an affricate, the two sounds were merged into the right syllable; 2) where two nasals were next to each other, they were

merged into the right syllable except when a fault-like boundary was discernable between the two nasals. As regards Italian: /sC/ clusters were treated as heterosyllabic clusters given the general tendency to shorten the duration of preceding vowels ([29], [30] and [31]). Long consonants were treated as the onset of the following syllable since no discontinuities occurred in the signal (fig. 2).

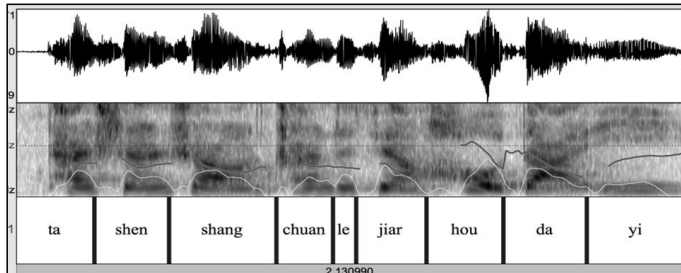


Figure 1 – An example of speech segmentation in Chinese

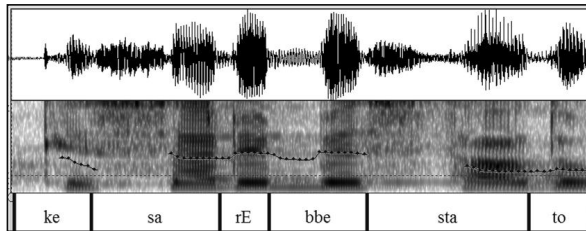


Figure 2 – An example of speech segmentation in Italian

From the syllable tier, we calculated raw and normalized duration-based metrics:

- ΔS : standard deviation of syllable duration
- r-PVI-S: averaged difference in duration between consecutive syllables
- VarcoS: normalized index of variability of syllable duration
- n-PVI-S: averaged of the mean difference in duration between consecutive syllables.

We also measured the average f0 across each syllable and, based on interval-based metrics, we devised the following F0 metrics: $\Delta S.F0$, VarcoS.F0 and r-PVIS.F0 and n-PVIS-F0. In addition, we developed combined metrics based on both duration and F0. For each syllable we measured the duration in seconds and mean F0 in Hz. We considered syllable duration (D) and mean syllabic F0 ($f0$) as two elements of a single vector ($mF0$). We calculated the standard deviation and the PVI of the norm of such vector ($= \sqrt{D^2 + mF0^2}$) sentence-wise, resulting in the following hybrid measures: Δ combined, Varco combined, rPVI combined and nPVI combined.

2.3 Data analysis and results

To test the significance of between-language variability (L1 Chinese, L2 Chinese and L1 Italian) on the three groups of rhythmic measures (duration, f0 and duration and F0 combined), we run a series of one-way ANOVA in SPSS with the values of the metrics as the dependent variables and the languages as the independent variable.

2.3.1 Duration-based measures

A significant effect of language was found for all metrics, with the exception of N-PVI-S. ΔS : [$F(2,42) = 21.37, p < 0.0001$]; VarcoS: [$F(2,42) = 9.14, p < 0.001$]; r-PVI-S: [$F(2,42) = 11.96, p < 0.0001$]; n-PVI-S: [$F(2,42) = 1.233, p = 0.302$] (fig. 3).

Post hoc tests (Tukey HSD) conducted on the three metrics where the main effect of language was found indicated that:

- On ΔS : Chin_L2 was significantly higher than Chin_L1 ($p < 0.001$) and Ita_L1 ($p < 0.001$), whereas the difference between the latter two groups was not significant ($p = 0.585$) (fig. 3a)
- On VarcoS: Chin_L2 was not significantly lower than Chin_L1 ($p = 0.221$), whereas Ita_L1 was significantly higher than both Chin-L1 ($p = 0.037$) and Chin-L2 ($p < 0.001$) (fig. 3b)
- On r-PVI-S: Chin_L2 was significantly higher than Chin_L1 ($p < 0.001$) and Ita_L1 ($p < 0.001$), whereas the difference between the latter two groups was not significant ($p = 0.757$) (fig. 3c).

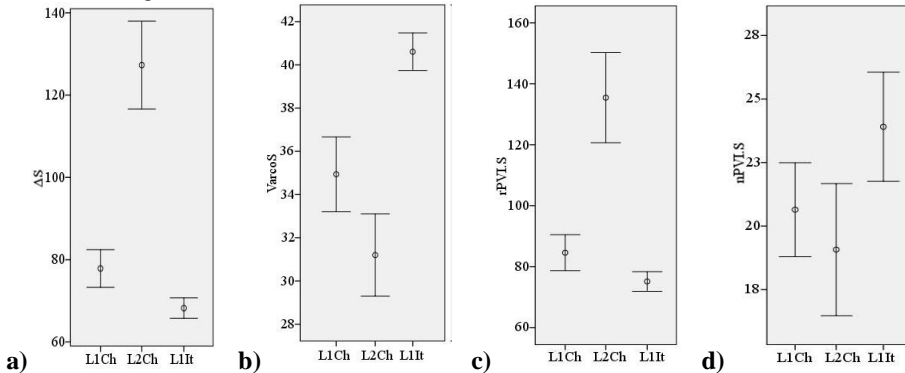


Figure 3 – Means and standard errors of metrics based on syllable duration per language group (L1Chin, L2Chin and LIIt); ΔS (a), VarcoS (b), rPVIS (c), nPVIS (d).

2.3.2 F0-based measures

A significant effect of language was found for all metrics, with the exception of VarcoF0. $\Delta S.F0$: [$F(2,42) = 3.496, p = 0.039$]; VarcoS.F0: [$F(2,42) = 1.048, p = 0.36$]; r-PVIS.F0: [$F(2,42) = 11.9, p < 0.0001$]; n-PVIS.F0: [$F(2,42) = 6.741, p = 0.003$] (fig. 4).

Multiple comparisons (Tukey HSD) conducted on the three metrics where the main effect of language was found revealed that:

- On $\Delta S.F0$: Chin_L2 was not significantly different either from Chin_L1 ($p = 0.256$) nor Ita_L1 ($p = 0.567$), whereas the difference between the latter two groups was significant ($p = 0.031$) (fig. 4a).
- On r-PVI-S: Chin_L2 was not significantly different from Chin_L1 ($p = 0.991$) but was significantly higher than Ita_L1 ($p < 0.001$). The difference between the latter two groups also was significant ($p < 0.001$) (fig. 4c).
- On n-PVIS.F0: the differences between Chin_L2 and Chin_L1 as well as between Chin_L1 and Ita_L1 were insignificant ($p > 0.05$), whereas Chin_L2 was significantly higher than Ita_L1 ($p = 0.002$) (fig. 4d)

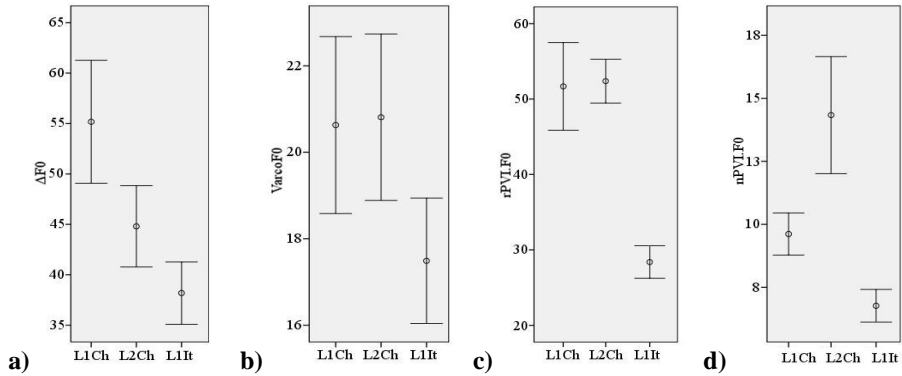


Figure 4 – Means and standard errors of metrics based on mean F0 per language group (L1Chin, L2Chin and L1It); $\Delta F0$ (a), VarcoF0 (b), rPVIF0 (c), nPVIF0 (d).

2.3.3 Combined measures (duration and F0)

A significant effect of language was found for all metrics: Δ combined [$F(2,42) = 24.72, p < 0.0001$]; Varco combined [$F(2,42) = 7.649, p = 0.001$]; rPVI combined [$F(2,42) = 5.715, p = 0.006$]; and nPVI combined [$F(2,42) = 5.238, p = 0.009$] (fig. 5)

Post hoc multiple comparisons (Tukey HSD) showed that:

- for Δ combined: Chin_L2 was significantly different from Chin_L1 and Ita_L1 ($p < 0.01$), whereas the differences between the latter two groups were insignificant ($p = 0.381$) (fig. 5a)
- for Varco combined: Chin_L2 was significantly different from Chin_L1 and Ita_L1 ($p < 0.01$), while Ita_L1 was not significantly higher than Chin_L1 ($p = 0.988$) (fig. 5b)
- for rPVI: Chin_L2 was not significantly different from Chin_L1 ($p = 0.198$), whereas the differences with Ita_L1 were significant ($p < 0.01$). The differences between Chin_L1 and Ita_L1 were insignificant ($p = 0.244$) (fig. 5c)
- for nPVI: Chin_L2 was significantly higher than both Chin_L1 ($p = 0.016$) and Ita_L1 ($p = 0.026$), while Ita_L1 and Chin_L1 were not significantly different ($p = 0.977$) (fig. 5d)

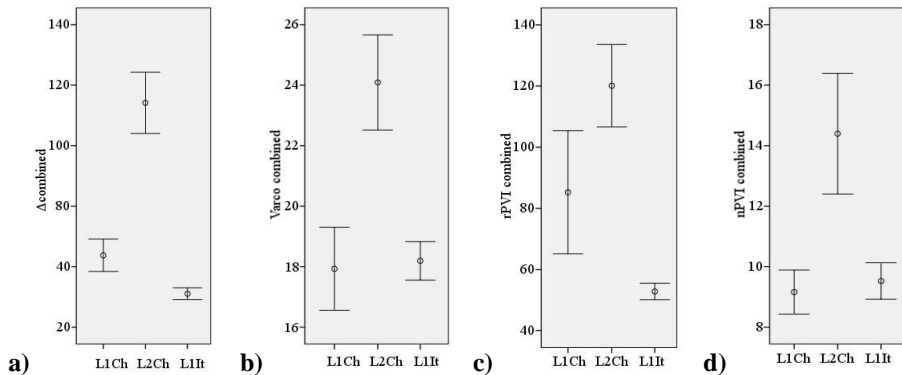


Figure 5 – Means and standard errors of combined metrics (duration and F0) per language group (L1Chin, L2Chin and L1It); Δ combined (a), Varco combined (b), rPVI combined (c), nPVI combined (d).

3 Discussion and conclusion

The overall results of the study show that metrics taking into account the variability of syllable duration and F0 separately do not distinguish consistently between L1 and L2 Chinese. The results largely depend on the acoustic parameter (syllable duration or mean syllabic F0) and on the metric examined (table 1: rows 1-6, columns: 3-4).

Regarding the durational characteristics of syllabic intervals (table 1: light grey columns), L2 Chinese overshoots the values of their native and target language in terms of ΔS and r-PVIS but it approximates the target pattern when the values are normalized for speech rate (Varco and nPVI_syllable duration). Given that the former two metrics have been repeatedly proven to be highly influenced by the speakers' speech rate [3], [4] the higher variability of syllabic intervals in L2 Chinese speech can be thus explained by the slower speech rate of L2 Chinese speakers. These results coupled with studies showing that L2 learners displayed increased consonantal and vocalic variability compared to the native speakers ([14],[15], [19]).

As regards the F0-based metrics (table 1: white columns), the differences between the two groups did not reach the significance level, suggesting that L2 Chinese learners approximate the native group in the production of lexical tones. One possible explanation to the nearly-native like performance of L2 Chinese speakers in F0-based metrics can be found in the extensive tone training they had done while studying Chinese at the University. There is evidence indeed that the L2 learners' ability to perceive Mandarin tones can improve significantly after perceptual tone training and this improvement transfers also to the production domain [32], [33].

Table 1: Metrics that significantly distinguish the three language pairs.

	Row	L2 Chin vs L1Chin	Chin_L2 vs Ita_L1	Ita_L1 vs Chin_L1
Duration-based measures	1.	ΔS	ΔS	-
	2.	rPVI-S	rPVI-S	-
	3.	-	<i>VarcoS</i>	<i>VarcoS</i>
F0-based measures	4.	-	-	$\Delta S.F0$
	5.	-	<i>rPVI.F0</i>	<i>rPVI.F0</i>
	6.	-	nPVI.F0	-
Combined measures	7.	Δ combined	Δ combined	-
	8.	-	r-PVI combined	-
	9.	Varco combined	Varco combined	-
	10.	nPVI combined	nPVI combined	-

A total different picture is displayed when L1 and L2 Chinese speakers are compared by measures combining duration and F0 (table 1, dark grey columns). Three out of the four combined measures gave values for L2 learners that were significantly higher from both L1 Chinese and L1 Italian (fig. 5). Interpreting these outcomes in the light of interlanguage hypothesis [34], it seems that pre-intermediate learners of Chinese have created 'a new rhythmic system' that is neither L1-like nor L2-like.

Although the results seem to be in favor of an approach that measures L2 speech rhythm combining two acoustic correlates of prominence, more research on different L1/L2 pairs is needed to verify whether these metrics are actually reliable for quantifying rhythmic

differences between L1 and L2 speech or among learners at different proficiency levels. Moreover, given that speech rhythm is a perceptual phenomenon, it would be important to verify to what extent combined measures correlate with listeners' ratings of perceived accentedness.

4 References

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