

Speech intelligibility measurement

A latent variable approach on utterances' transcriptions

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Abstract

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1 Introduction

Intelligible speech can be defined as the extent in which the elements in a speaker’s acoustic signal, e.g. phonemes or words, can be correctly recovered by a listener [9, 11, 18, 19]. Intelligible spoken language carries an important societal value, as its attainment requires all core components of speech perception, cognitive processing, linguistic knowledge, and articulation to be mastered [9]. In that sense, *speech intelligibility* is considered a milestone in children’s language development, and more practically, it is qualified as the ultimate checkpoint for the success of speech therapy, and the ‘gold standard’ for assessing the benefit of cochlear implantation [4].

Multiple approaches can be taken to quantify *speech intelligibility* [1, 2, 7, 10], but among them, *objective rating* methods on stimuli recovered from spontaneous speech tasks have received special attention [2, 10]. In objective rating methods, listeners transcribe children’s utterances orthographically (or phonetically), and use such information to construct an intelligibility score. The construction of the score can be done in several ways, e.g. counting the number of (un)intelligible syllables or words in the utterances [7, 12], or calculating the transcriptions’ entropy, a measure that expresses the degree of (dis)agreement in the data [2, 16]. In that sense, the method tries to infer intelligibility, from the extent in which a set of transcribers can identify the words contained in multiple utterances [2].

As the literature suggests, objective rating procedures produce more valid¹ and reliable² scores than any other available procedure [2, 6], as the method does not hinge in the use or production of a *subjective rating scale*, i.e. a scale based on a personal perception of the child’s intelligibility. Moreover, the previous advantages are further emphasized by the use of stimuli gathered from spontaneous speech tasks, as they have a greater level of ecological validity, especially compared to contextualized utterances or reading at loud tasks [7, 5].

However, although the literature is clear on the method’s benefits to measure *speech intelligibility* [1, 2, 10], we notice the statistical approaches used to model such data still face three important issues, and these come to the detriment of the measurement procedure’s sophistication.

First, as previous paragraphs reveal, the intelligibility scores are ‘complex’ in nature, however, such ‘complexity’ is rarely fully considered in the statistical modeling procedure. The problem with the later is that, because the data does not fulfill the typical assumptions, e.g. normality, its analysis under such models might lead us to erroneous conclusions [citation]. On the one hand, outcomes such as the number of (un)intelligible words are discrete, while the entropy scores are continuous in nature. In addition, there is the consideration that both measures are constraint in specific bounds, i.e. the number of (un)intelligible words cannot be negative, while the entropy scores are in the bounds between zero and one. Finally, given the measurement procedure’s nature, the scores are produced in a clustered manner, i.e. we observe several score measurement per child. Considering all of the above, it is clear the modeling requires some adjustments to account for all of these nuances, in favor of proper statistical inferences.

So far the literature shows the applied statistical procedures have always assumed ‘normality’, examples of this can be seen in Boonen et al. [2], Flipsen and Colvard [8] and Hustad et al. [10]. In addition, some papers in the literature have even used multilevel modeling to deal with the clustered nature of the data, an example of this can be found in Boonen et al. [2]. However, to the authors knowledge, no paper have dealt with all of the data ‘complexity’ at once, which leads us to believe that, by using more sophisticated statistical models we could improve our statistical inferences.

Second, coupled with the previous, no attempt on constructing an actual intelligibility score is made, and therefore, the modeling procedure is applied on surrogate measures, like entropy, but not directly on what it is intended to be measure. For instance, Flipsen [7] uses

¹validity is understood as the extent to which scores are appropriate for their intended interpretation and use [14, 17].

²reliability is though as the extend to which a measure would give us the same result over and over again [17], i.e. measure something, free from error, in a consistent way.

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