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**Investigating potential acoustic correlates of sonority:
Intensity vs. periodic energy**

Bachelorarbeit an der Philosophischen Fakultät der Universität zu Köln
im Fach Linguistik und Phonetik

vorgelegt von

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Köln, 10.08.2020

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1 Introduction

Spoken language consists of speech sounds that combine into syllables, and further on, into words and phrases. Within syllables, speech sounds cannot be combined completely arbitrarily. For example, the sequence /plid/ is considered a syllable in, e.g., the English language, but /lpid/ is not. In order to describe what sounds can be combined in which way, the concept of sonority is consulted. Generally speaking, sounds with a high sonority (mostly vowels) form the nucleus, i.e. the centre, of a syllable. They can (but do not have to) be surrounded by other sounds (mostly consonants) with a lower sonority than the nucleus. From the syllable onset, i.e. its beginning, to its nucleus, sonority is expected to rise. Conversely, from nucleus to coda, i.e. the syllable's end, sonority is expected to fall. Regarding the example above, /l/ is considered more sonorous than /p/, which is why /plid/ forms a syllable, in contrast to /lpid/. The same applies to codas: /help/ is considered a syllable, whereas /hɛpl/ is not.

The assignment of sonority ranks to speech sounds primarily grounds on linguistic observations on how speech sounds can be combined in different languages. Still, there is no agreement on whether sonority has a physical, measurable basis. Some researchers even deny the concept of sonority entirely. In this thesis, I aim at contributing to the discussion by exploring possible acoustic correlates of sonority. To do so, I will first give an overview of the research on sonority (section 2). Following, I will report on an exploratory pilot study I conducted, describing and discussing the acoustic measurements in detail (section 3). In section 4, the thesis finishes with a conclusion and an outlook on further research.

The data and the code from this study can be found at https://osf.io/w5cje/?view_only=f3b7aa80fcc940aba83ddf21512efe5c.

2 Overview of sonority

This section gives an overview on the subject of sonority. First, the scientific background is presented (subsection 2.1). Subsequently, I explain my decision to explore possible sonority correlates in the acoustic domain, taking into account recent findings (subsection 2.2).

2.1 Background

The discussion on the topic of sonority as a core principle behind syllables has quite a long history in linguistics, with Sievers (1893/1901) being the first to introduce it by the term *Schallfülle* ('sound fullness'). He states that every speech sound has a certain sonority value and that within a syllable, the sounds with a relatively low sonority (which Sievers links to loudness) mark its boundaries, and, in turn, very sonorous sounds form the syllabic nucleus. According to him, however, there is no uniform definition of the term 'syllable'. He describes them as subjective perceptual segmentations of speech, which are reported to be based on discontinuities in sonority. Thus, his work can be seen as an early approach to define syllables – an issue which until today has not been solved.

There has been an ongoing debate about the notion of sonority, regarding its definition, possible measurable correlates, or its existence altogether. Kawasaki (1982) sheds light on the missing definition of the idea of sonority. She presents a debate on whether the concept should be explained phonologically or phonetically. She argues for a phonetic approach, because phonological explanations were prone to circular reasoning.

Ohala (1992) holds a similar view. He completely rejects the idea of sonority due to a missing measurable correlate and the thereby inherent circularity in its definition: In his view, sonority only describes the preferable ordering of sounds in a syllable, which remains hard to define with any consistent criterion. Since the preference for some speech sound combinations cannot be accounted for by the sonority hierarchy and some relatively prevalent consonant clusters (such as /st-/ or /-kf/) even violate the hierarchy, he suggests to directly consult the physical parameters amplitude, periodicity, spectral shape and fundamental frequency (F0), 'which are well known and readily measured in the speech signal' (Ohala 1992: 325).

Clements (1990) argues that the fact that there is no physical definition for a concept does not make it useless for research. This was true for sonority as well as phonemes or syllables. He proposes a number of phonological features (sonorant, approximant, vocoid, syllabic) in order to explain the cross-linguistically similar sonority scales.

According to Parker (2011), sonority ‘can be defined as a unique type of relative, n-ary (non-binary) featurelike phonological element that potentially categorises all speech sounds into a hierarchical scale’ (Parker 2011: 1160). He states that, as the classes of sonority are mostly categorised based on manner of articulation, this is the closest correlate in traditional phonetic terms. This statement, however, is just of descriptive nature and has no explanatory value due to its circularity. Parker (2011) calls sonority ‘featurelike’ because it can cluster speech segments into several groups. On the other hand, he describes sonority as being ‘unlike most features’ because without exception every segment bears a certain value of it. In a meta-study (Parker 2017), he analyses 264 studies dealing with sonority by one means or another, with approx. 57 % of them indicating a significant role of sonority for explaining the results, even MRI analyses of brain functions.

In response to the points of criticism regarding the phonetic substance of sonority (Kawasaki 1982; Ohala 1992), Parker (2008) conducts a study investigating the correlation between sonority (more precisely, his scale given in (2)) and intensity, which results in a mean Spearman’s correlation of 0.91. In contrast, Albert and Nicenboim (2020) argue that sonority refers to pitch intelligibility: Vowels are the most sonorous speech sounds and can best convey pitch. In contrast, a voiceless obstruent is not a good carrier of pitch information, due to its minimal amount of periodic energy. So, pitch intelligibility in turn correlates with periodic energy, a subpart of the overall intensity of the acoustic signal (Albert et al. 2018).

Another issue concerning sonority is that a number of different scales have been proposed. Parker (2002), e.g., reports on over 100 sonority hierarchies of which some are language-specific and others claim to be cross-linguistically applicable. In this study, I will consult three sonority scales of which one scale, put forward *inter alia* by Clements (1990), is often considered the lowest common denominator, see (1). All sonority hierarchies depicted here start with the most sonorous group of segments, going down to the least sonorous.

(1) Vowels

Glides

Liquids

Nasals

Obstruents

On the basis of phonotactics and phonological processes, Parker (2008) proposes a very detailed sonority hierarchy (see (2)).

(2) Low vowels

Mid peripheral vowels (not ə)

High peripheral vowels (not i)

Mid interior vowels (ə)

High interior vowels (i)

Glides

Rhotic approximants (ɹ)

Flaps

Laterals

Trills

Nasals

Voiced fricatives

Voiced affricates

Voiced stops

Voiceless fricatives

Voiceless affricates

Voiceless stops

Albert and Nicenboim (2020) propose a hierarchy which is based on the periodic energy potential of the acoustic signal considering coarse differences in voicing and articulatory contact (see (3)). The scale collapses the first four groups from (1) into two groups and divides the group of obstruents in terms of voicing, ending up with one group less than (1).

- (3) Sonorant vocoids (e.g. glides, vowels)
- Sonorant contoids (e.g. nasals, liquids)
- Voiced obstruents (e.g. stops, fricatives)
- Voiceless obstruents (e.g. stops, fricatives)

Proceeding from sonority scales, there are various approaches of combinatorial rules for composing well-formed syllables, among them the *Sonority Sequencing Principle*, the *Minimum Sonority Distance* (Steriade 1982, Selkirk 1984) and the recently proposed *Nucleus Attraction Principle* (Albert & Nicenboim 2020). However, as this thesis does not deal with syllabic well-formedness, these principles are not explained in detail. If the reader is interested in a thorough overview on the literature of sonority, the publications of Parker (2002; 2017) are recommended.

In summary, there is a considerable amount of evidence that the notion of sonority can explain phonological concepts such as syllable structure. Yet, as long as there is no agreement on a phonetic basis for sonority, this explanation remains circular: Sonority describes syllable structure, but on the other hand, syllable structure is based on sonority (and other phonotactic observations). In order to break this circularity, sonority has to be defined on an independent, phonetic basis.

2.2 Why acoustics?

In this thesis, I examine acoustic recordings to search for possible measurable correlates of sonority. I will outline this decision in the following. As Ladefoged (1997) points out, both articulation and perception play an important role in phonetics as well in phonology, because language only works if it can be produced and understood properly. Acoustics combine both dimensions, as they represent the output of articulation and the input of perception. Thus, acoustic recordings are a very accessible way to examine both of them. Of course, it is a challenge to find out which acoustic parameter reflects which articulatory or perceptual aspect. In this thesis, however, I only investigate possible acoustic correlates of sonority without further discussing their link to articulation or perception.

To date, no clear picture has emerged on which parameter suits best as an acoustic correlate. Parker (2002) measured intensity, F1 frequency and segmental duration for English and Spanish, in addition to some articulatory dimensions. He concludes that intensity performs best for both languages, compared to all other acoustic and articulatory parameters. He confirms the high correlation of intensity with sonority indices in a second study for these two languages and, in addition, Cusco Quechua. Jany et al. (2007), in turn, confirm this finding for Egyptian Arabic, Hindi, Malayalam and Mongolian, even though some cross-linguistic differences are reported.

Gordon et al. (2012) examine the values of duration, maximum intensity, F1 values, total acoustic intensity, and total perceptual energy (filtered total acoustic intensity) for five languages (Hindi, Besemah, Armenian, Javanese and K^wak^wala). Their study is only concerning vowels and aims at finding a measurement that predicts the low sonority value of schwa. They do not find a value that consistently performs correctly for all languages and conclude that sonority might be a multi-dimensional variable, which is in line with the findings from Komatsu et al. (2002) for Japanese consonants.

Another reason why acoustics are studied often is the possible enhancement of speech technology, such as automatic speech recognition (ASR). Kawai and van Santen (2002), Nakajima et al. (2012) and Patha et al. (2016) go in this direction. Kawai and van Santen (2002) use the RMS power of filtered frequency bands to discriminate between speech sound classes and in this way detect syllable nuclei. Nakajima et al. (2012) follow a similar approach, conducting a factor analysis for acoustic power. They find that a factor around 1000 Hz correlates best with sonority. Both studies were conducted with English recordings. Patha et al. (2016) analyse recordings for the languages Telugu, Tamil and Hindi. They band-pass filter the speech signal between 500 Hz and 1700 Hz and report that the amplitude of the dominant resonance frequency works best for detecting syllable nuclei and boundaries.

Ladefoged (1997) and Heselwood (1998) suggest to analyse periodic energy for sonority. Periodic energy forms part of the intensity of the acoustic signal, but is limited to its (quasi-)periodic portion (Albert et al. 2018). It could serve as a better sonority correlate than intensity: For example, the fricative /s/ might have a high intens-

ity because of the friction, but still a low sonority value, due to the missing periodic energy.

There has not been much research on the link between periodic energy and sonority. Llanos et al. (2015) compare intensity and Shannon entropy for English. Shannon entropy is akin to detecting periodic energy: a higher periodicity in the signal leads to a higher Shannon entropy. They report that Shannon entropy is a better correlate for their proposed hierarchy (vowel > approximant > nasal > fricative > affricate > stop) than intensity. It is notable that their database consists of 6300 sentences from 630 speakers. Albert and Nicenboim (2020) conduct a syllable count task and conclude that periodic energy (integrated over time) serves best as foundation for a sonority model to predict their experiment's results. Therefore, they propose the periodicity-based Nucleus Attraction Principle as a new model of sonority.

Due to the given evidence, I view acoustics as a promising chance to explore the phonetic nature of sonority. In the course of this thesis, I will explore the correlation between different measurements for intensity and periodic energy and three distinct sonority scales.

3 Study

In order to investigate possible sonority correlates, I analysed consonants from acoustic recordings from a pilot experiment conducted at the Institute for Linguistics – Phonetics, University of Cologne.

For every consonant, I examined various measurements of intensity and periodic energy so as to cover all main possibilities to measure a continuous curve: minimum and maximum peaks, sums and averages. By this, I wanted to eliminate the possibility that the results of this exploratory study are biased due to a specific measuring technique.

In order to conduct measurements on speech sounds, the audio stream has to be segmented. However, phones are not produced sequentially, but with a considerable, variable overlap (e.g. Bell-Berti & Harris 1981, Browman & Goldstein 1992). Therefore, audio segments are inevitably influenced by the neighbouring segments and cannot provide for a clear-cut distinction. Nevertheless, for an acoustical analysis of single speech sounds, a segmentation has to be made based on predefined criteria.

These criteria and the experimental setup will be explained in detail in subsection 3.1. Further on, I will report on the results of this exploratory study in subsection 3.2, followed by a more detailed analysis (subsection 3.3) and the discussion (subsection 3.4).

3.1 Method

In this section, I will outline the setup of the experiment that produced the recordings I studied (subsection 3.1.1). After that, I will elaborate on my analysis procedure, concerning the measurements I conducted (subsection 3.1.2) and the statistical calculation (subsection 3.1.3).

3.1.1 Experimental setup

The experiment was carried out with five participants (1 female, 4 male). Three of them had German as a native language, one Dutch and one Hebrew. All speakers had grown up monolingually and were trained phoneticians. Participants were asked to read out a carrier phrase in their native language with one alternating target pseudo-word in non-final position. For German, the phrase was ‘Ich möchte [...] kaufen’ (‘I would like to buy [...]’). In Dutch, the phrase was ‘Wij kunnen [...] kopen’ (‘We can buy [...]’), and for Hebrew, it read ‘הוא רואה [...] כרגע’ (/ʔ(h)u ʁo.ˈe [...] ka.ˈʁe.ga/ ‘He is seeing [...] at the moment’).

The target word had the form of /aˈCa/, an alternating consonant surrounded by the vowel /a/. Each of the 17 alternating consonants was produced and recorded once per participant. I limited my study to bilabial, labiodental and alveolar consonants as well as glides, so four consonants (/k/, /g/, /x/ and /ʃ/) were excluded. For a list of the 13 consonants that went into this study, see Table 1 in subsection 3.1.3.

3.1.2 Measurements

The recordings were edited by extracting the target words from the carrier phrases and adding a 100 ms padding before and after. Each file was then normalised to a loudness target of -20 LUFS (Loudness Units relative to Full Scale, ITU-R 2015). The segmental annotation was done with Praat (Boersma & Weenink 2020), the statistical analysis was carried out using R (R Core Team 2020). The extraction of in-

tensity and periodic energy was based on a workflow for Praat and R (Albert et al. 2020), building upon methods from Albert et al. (2018).

As outlined before, a segmentation of the speech stream is required for this analysis, although segments are always influenced by neighbouring segments. It is therefore important to describe the criteria for segmentation. Primarily, I segmented the stimuli into speech sounds on the basis of the spectrogram and oscillogram. This is relatively easy for voiceless sounds, because the consonants are surrounded by (voiced) vowels and the oscillogram and the spectrogram show obvious differences. For voiced sounds, these differences are not that clear, though in most cases discoverable in terms of visible changes in acoustic patterns (see Figure 1 as an example). Glides, being the group of consonants with least constriction in the vocal tract (and highest sonority indices), are most difficult (see Figure 2). When in doubt, I decided based on hearing impression. In a separate annotation step, I subdivided the stops into closure and burst phase (motivation for this approach following below). For that, I put the segment boundary right before the visible burst in the spectrogram, preferably at a zero-crossing point in the oscillogram, if possible.

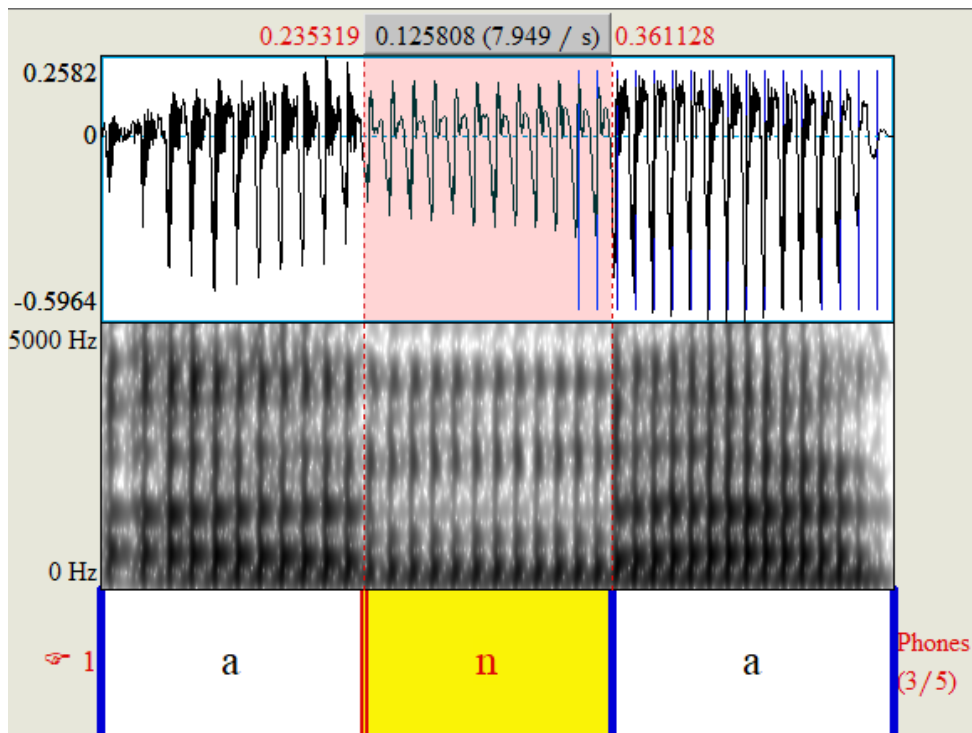


Figure 1: Oscillogram, spectrogram and segmental annotation for /a'na/. The differences between nasal and vowels are not very substantial; however, the nasal exhibits a less complex waveform and a slightly different formant structure. Screenshot from Praat.

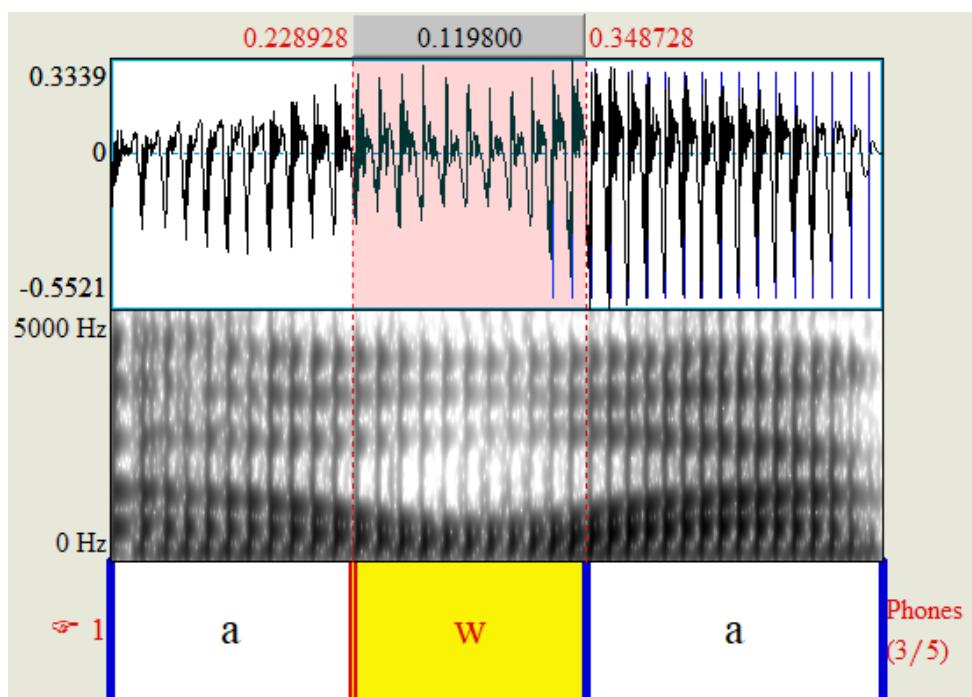


Figure 2: Oscillogram, spectrogram and segmental annotation for /a'wa/. Although there is a visible change in formants and waveform, this change occurs gradually and no distinct boundary can be determined. Screenshot from Praat.

The variables I analysed are intensity and periodic energy. Intensity refers to the power of the whole signal, whereas periodic energy is the power of its periodic components. It is important to keep in mind that dealing with continuous acoustic measurements means dealing with a curve, rather than discrete values. A curve always comes with two dimensions – in this case, power and time. There are several valid measurements that account for these two dimensions: peaks, summing and averaging. I will implement all of them so as to explore which measuring technique suits best for this study.

Some of the measurements are sensitive to segmentation. In natural language, we cannot determine one clear point in time where one segment ends and the other begins. However, for studies like this, which link continuous acoustic variables to discrete categorical representations of segments, the segmentation of the audio stream is an inevitable procedure. The effects of this methodological problem can be seen in the following example. Figure 3 shows the curve of smoothed periodic energy for the stimulus /a'pa/. As is visible, the stop contains some of the periodic energy of the surrounding vowels, leading to a high maximum in periodic energy and also influencing sum and average. Only the minimum peak would not be changed in this case.

In the following subsections, I will elaborate on the measurements I conducted. Subsection 3.1.2.1 deals with peaks, whereas subsection 3.1.2.2 addresses summing and averaging. Each measurement was carried out for both variables, intensity and periodic energy.

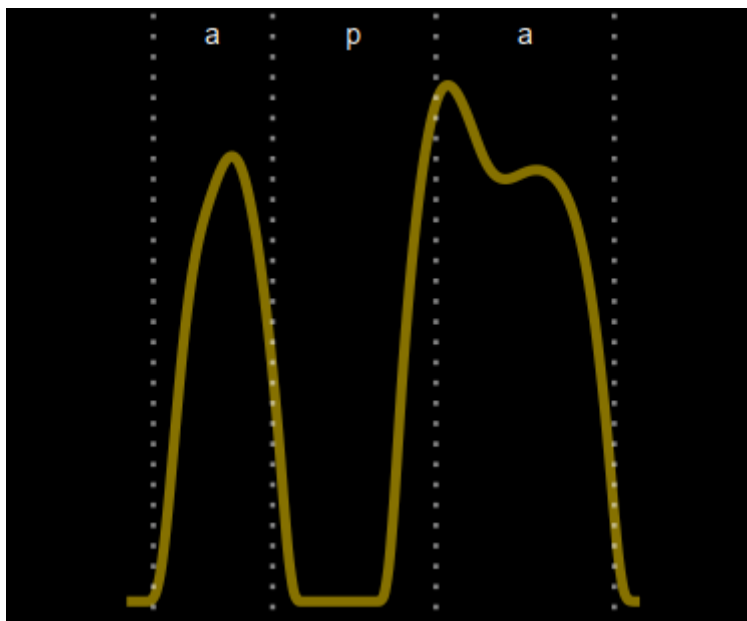


Figure 3: Diagram of the recording /a'pa/ (x axis: time, y axis: periodic energy) with its segmentation. Because no single segment border can be established naturally, influences by surrounding segments are visible. Although being a voiceless stop, /p/ contains some periodic energy of the previous and following /a/. Diagram created by means of Praat & R workflow (Albert et al. 2020).

3.1.2.1 Peaks

Peak values are a simple measuring technique: In our case, for a given time window, the minimum or maximum value of the examined variable (intensity/periodic energy) is identified. Since this is an exploratory study, I analysed both minimum and maximum values.

Parker (2008) measured the intensity minimum for consonants and the maximum for vowels. He justifies this decision by the fact that this would eliminate segmentation issues, because for consonants, the minimum intensity is likely to occur in the centre of the audible consonant and not at the boundaries. The same applies to vowels and their maximum. This way, the exact segmentation is not important. Apart from this basic minimum value, Parker (2008) comes up with a special measurement: Because aspirated stops often exhibit two local minima (one during closure and one during burst phase), Parker (2008) averages these two minima in such cases. In order to review this measurement, I set up a scale with simple minima for all sounds except

aspirated stops. For the latter, I selected the complex measurement with averaged minima (see Figure 4). Three of the five participants (the German speakers) aspirated their voiceless stops. Furthermore, since not only aspirated, but all stops can be subdivided into closure and burst phase, I added a third scale of minima: This scale contains the complex minima for all stops by all speakers and the simple ones for all other sounds.

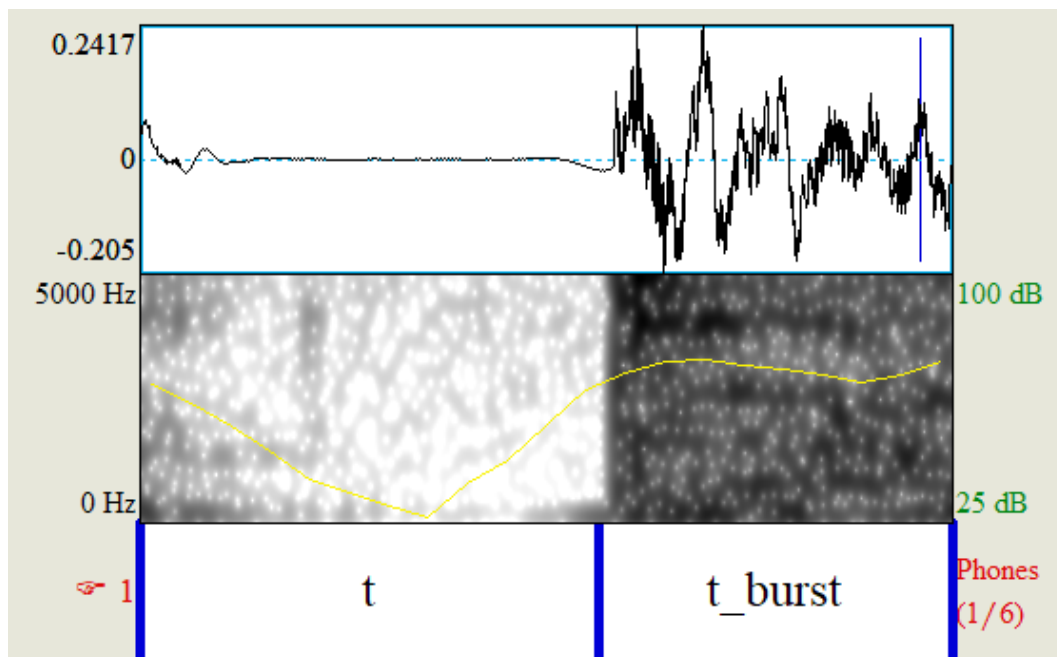


Figure 4: Oscillogram and spectrogram for /t/ with aspiration, annotated for closure and burst phase. The curve in the spectrogram depicts intensity, with two visible local minima. For the simple minima measurement, the overall minimum (contained in the closure) is considered. For the complex measurement, the minima of closure and burst are averaged. Screenshot from Praat.

Up to now, we have four different peak-based measurements each for intensity and periodic energy: maximum, minimum, minimum with complex minimum for aspirated stops, and minimum with complex minimum for (all) stops. In order to account for possible variation in the overall intensity of the pronounced utterance or word, these scales were additionally normalised with regard to the respective maximum value of the following vowel (which was of the same type in all utterances), leading to eight scales per variable.

3.1.2.2 *Summing and averaging*

Peak values only consider one point in time: The moment when the maximum or minimum of the analysed variable is reached. All other parts of the time window are ignored by this technique. In contrast, summing and averaging take the whole time window into account: A sum, also referred to as the area under the curve (AUC), sums up every value of intensity/periodic energy for all points in time of the segment. An average divides this number by the segment's duration, resulting in a mean value of the examined variable.

For example, Albert and Nicenboim (2020) measure the area under the curve (AUC) for periodic energy, which corresponds to the integration of the power and time. This has two main effects. First, in contrast to peak values, an integral covers not only one specific point of time, but the whole segment. This can be demonstrated with Figure 3: The segment /p/ has a high maximum value of periodic energy which is almost as high as the following vowel's maximum. However, the AUC of the consonant is visibly smaller than that of the neighbouring vowels, because on the whole, /p/ exhibits a little amount of periodic energy, and its high maximum can be attributed to the following /a/. Second, the duration of a segment is assumed to be possibly associated with sonority, and measuring the AUC takes this duration into account: A short segment has a lower AUC value than a longer segment with a similar average value for the regarded variable (intensity/periodic energy). Therefore, I included the sum (AUC) of the consonantal segments as another scale.

As stated above, a mean value takes the whole duration of the examined segment into account, but in contrast to a sum, it is normalised. If we compare two recordings of the same consonant, e.g., a nasal, produced with approximately the same power but differing in duration, the sum of power and time in these recordings would differ considerably, but its mean value would stay (roughly) the same.

Gordon (2004) introduces the measure of 'total perceptual energy' in a work on prosodic weight. This involves averaging of acoustic energy and normalisation in relation to a vocalic anchor, after which the result is transformed into a scale that tries to estimate perceived loudness. Finally, this is again multiplied by the duration of the segment. This measurement, though, is not conducted in this study. First, it involves

many researcher degrees of freedom. Second, it mixes summing and averaging, and part of this study is to explore the differences between these two calculations.

Similar to my procedure with the peak values, I set up two additional scales in which sum and mean, respectively, are normalised in relation to the corresponding value of the following vowel.

3.1.3 Statistical calculation

Taken together, this study considers 12 scales per variable and therefore 24 scales in total. For all these scales, correlation values were computed with the three sonority hierarchies from section 2.1. Table 1 shows the consonants from this study and their rank in the respective hierarchies.

class		consonant	(1) Clements (1990)	(2) Parker (2008)	(3) Albert & Nicenboim (2020)
sonorants	glides	j	4	7	4
		w	4	7	4
	liquids	l	3	6	3
	nasals	m	2	5	3
		n	2	5	3
obstruents	voiced fricatives	v	1	4	2
		z	1	4	2
	voiced stops	b	1	3	2
		d	1	3	2
	voiceless fricatives	f	1	2	1
		s	1	2	1
	voiceless stops	p	1	1	1
		t	1	1	1

Table 1: The analysed consonants and their scoring in the investigated sonority hierarchies (cf. section 2.1).

The depicted scales were chosen to reflect prototypical sonority hierarchies: The scale by Clements (1990) is a traditional hierarchy with distinctions between the classes, but no further distinction within the class of obstruents. Parker (2008), on the

other hand, divides the latter into four subclasses. This hierarchy can thus be seen as a maximally divided scale based on the traditional approach. Scale (3) by Albert and Nicenboim (2020) is, as mentioned above, based on the periodic energy potential of the consonants and thus reflects a new approach to sonority.

For each of the afore-mentioned 24 measured scales, the mean values per consonant were calculated. I then computed correlations between each measured scale and each sonority hierarchy, ending up with 72 correlation values. The applied rank correlation coefficient was Spearman's ρ , as this is suitable for comparing two data series of different types (Parker 2008). In this case, the measured values form (continuous) ratio scales, which means that between two values, a ratio can be established (for instance, 1.5 is the double of 0.75). In contrast, the sonority hierarchies have (discrete) ordinal values. This means that a consonant with sonority rank 3 has a higher sonority than one ranked 2, which in turn is higher than rank 1. However, no ratio can be calculated: It is not determined whether the distance between 2 and 3 is the same as between 1 and 2, and whether 2 is the double of 1. The chosen correlation coefficient takes these differences into account.

For a more detailed analysis, I selected eight scales that I considered deserving further investigation and calculated Tukey groups. These groups give an insight into which consonants differ significantly from each other in one measured scale.

3.2 Results

In this section, I will report on the correlation results. First, I will examine the three types of minimum measurements (see Table 2) and discard two of them from further analysis, reducing the number of scales. Then, I will compare the remaining results with each other (see Table 3). The legend printed below explains the variable names for both tables.

Legend:

int	intensity
pe	periodic energy
max	peak maximum value
min	peak minimum value
min_stops	peak minimum value (for stops: averaged minimum between closure and burst)
min_asp	peak minimum value (for aspirated stops: averaged minimum between closure and burst)
sum	sum (area under the curve) of whole consonant's duration
mean	mean value of whole consonant's duration
rel	value relative to the following vowel's respective value (note: minimum peaks are normalised in relation to the following vowel's <i>maximum</i>)

minimum measurement	(1) Clements (1990)	(2) Parker (2008)	(3) Albert & Nicenboim (2020)	average
int_min	0.761	0.881	0.880	0.841
int_min_rel	0.770	0.878	0.880	0.843
int_min_stops	0.761	0.837	0.880	0.826
int_min_stops_rel	0.770	0.845	0.880	0.832
int_min_asp	0.761	0.881	0.880	0.841
int_min_asp_rel	0.770	0.878	0.880	0.843
pe_min	0.771	0.871	0.892	0.845
pe_min_rel	0.781	0.879	0.892	0.851
pe_min_stops	0.762	0.816	0.881	0.820
pe_min_stops_rel	0.771	0.813	0.881	0.822
pe_min_asp	0.762	0.838	0.881	0.827
pe_min_asp_rel	0.771	0.846	0.881	0.833

Table 2: Correlation values for the different types of minimum measurements and the three sonority hierarchies, rounded to the third decimal place. The last column shows the average correlation value for all three hierarchies. The variables whose names are printed in bold were kept for further comparison with the other measurements (maximum, sum and mean).

As written above, I measured three types of minima. The first measurement, which is referred to by *min* in Tables 2 and 3, is simply the lowest value of the investigated variable (intensity or periodic energy) for each segment – analogous to the maxima. Parker (2008) uses a different measuring technique for aspirated stops: He separates them into closure and release phase and calculates the average of the minima of both. I followed that approach in the measurement *min_asp*: for most sounds, it still contains the simple minimum value, but for aspirated stops, it is the averaged minimum. Finally, as every stop can be subdivided into closure and release, I added a third measurement (*min_stops*) with the averaged minimum for *all* stops and the simple minimum for all other sounds.

It is important to note that only the German speakers (three out of five) aspirated their voiceless stops. As a consequence, *min_asp* contains the averaged minimum only for the voiceless stops produced by the German speakers. In contrast, *min_stops* treats all speakers the same way. However, this means that many rather short release phases of unaspirated stops come into analysis. Their minima are possibly located at the segmental boundaries, in contrast to the longer aspirated bursts that tend to exhibit minima in their centre.

All measurements were conducted in an absolute and a relative (*_rel*) version. The absolute version corresponds to the value obtained directly from the audio signal of each segmented consonant. For the relative version, the consonant's value is divided by the following vowel's respective value. This means that, e.g., *int_sum_rel* is the summed intensity of a consonant divided by the following vowel's summed intensity. As an exception, in the case of the minima, the consonant's value is divided by the following vowel's maximum, not its minimum.

Table 2 shows the correlation values between the minima and the investigated sonority hierarchies, plus a mean value of all three hierarchies for each measurement. Comparing the results with each other, it becomes apparent that for all hierarchies, the complex measurements (*min_stops* and *min_asp*) score the same or worse than the simple measurement. This finding is the same for both intensity and periodic energy, and both the absolute and relative measurements. Thus, the simple minima are kept for further comparison, especially since they make less assumptions than the complex ones.

measurement	(1) Clements (1990)	(2) Parker (2008)	(3) Albert & Nicenboim (2020)	average
int_sum	-0.101	-0.465	-0.441	-0.336
int_sum_rel	0.041	-0.357	-0.330	-0.215
int_max	0.409	0.238	0.421	0.356
int_max_rel	0.758	0.629	0.752	0.713
int_mean	0.789	0.862	0.908	0.853
int_mean_rel	0.780	0.853	0.908	0.847
int_min	0.761	0.881	0.880	0.841
int_min_rel	0.770	0.878	0.880	0.843
pe_sum	0.808	0.867	0.922	0.866
pe_sum_rel	0.808	0.867	0.922	0.866
pe_max	0.817	0.720	0.823	0.787
pe_max_rel	0.808	0.778	0.877	0.821
pe_mean	0.789	0.873	0.908	0.857
pe_mean_rel	0.770	0.845	0.894	0.836
pe_min	0.771	0.871	0.892	0.845
pe_min_rel	0.781	0.879	0.892	0.851

Table 3: Correlation values for all measurements (except the complex minima) and the three sonority hierarchies, rounded to the third decimal place. The last column shows the average correlation value for all three hierarchies. For every hierarchy, the measurements that scored best are printed in bold and underlined. The second best measurements are printed in bold without underlining. The variables whose names are printed in bold went into further analysis.

Inspecting Table 3 with all the correlation values except the complex minima, it can be noted that most measurements still score similarly. Three measurements are standing out in a negative way: summed intensity, both absolute and relative (*int_sum*, *int_sum_rel*), and (absolute) maximum intensity (*int_max*). The intensity sum values are the only ones to exhibit very low and negative correlations. The maximum intensity values have very low positive correlations. Interestingly, when normalising the consonant’s maximum intensity by the following vowel’s maximum intensity (*int_max_rel*), the correlation increases considerably (average: 0.356 vs.

0.713). This is the only case where the difference between a measure's absolute and relative version is that high.

Still, even *int_max_rel* exhibits, on average, considerably lower correlation values (0.71) than the remaining 20 measures, which range between an overall average of 0.79 and 0.87. For hierarchy (1), the correlation values of those 20 measurements range between 0.75 and 0.82. In the case of (2), they range from 0.72 to 0.88, and for (3), a range spanning from 0.82 to 0.92 can be observed.

3.3 Analysis

For further investigation, I calculated Tukey groupings to get a better insight into the single scales. Tukey groups indicate for each measurement which consonants differ significantly from each other. These groups can then be compared with the proposed sonority hierarchies. Because a detailed comparison of all 16 scales would go beyond the scope of this thesis, I chose to examine only one version (absolute or relative) of each measurement, ending up with eight variables.

In the case of *int_max*, the relative version scores much better than the absolute one (on average, 0.713 vs. 0.356). For the other measures, the correlation values of their absolute and relative version differ only slightly. Thus, in order to provide for the best possible comparability, I examined the relative version of each measurement.

3.3.1 Peaks

3.3.1.1 Minima

consonant	Minimum intensity	Tukey groups (intensity)	Minimum periodic energy	Tukey groups (periodic energy)
j	0.947	dfg	0.701	b
w	0.939	dfg	0.664	b
l	0.954	fg	0.737	b
m	0.952	efg	0.731	b
n	0.963	g	0.791	b
v	0.832	cd	0.142	a
z	0.844	cf	0.236	a
b	0.840	cf	0.105	a
d	0.837	cde	0.204	a
f	0.748	bc	0.000	a
s	0.816	c	0.000	a
p	0.605	a	0.000	a
t	0.633	ab	0.000	a

Table 4: Relative minimum values of intensity and periodic energy for each consonant with their Tukey groups. The displayed values are averaged across all speakers.

Table 4 shows the mean values and Tukey groups for the examined minima. It can be noted that the periodic energy measurements clearly subdivide the consonants into obstruents and sonorants. No other subdivision is apparent from the Tukey groups: Voiceless and voiced obstruents do not differ significantly from each other, nor do, e.g., nasals and glides. When looking at the mean values, however, voiceless obstruents exhibit no periodic power at all, in contrast to all other consonants.

Intensity shows more subdivisions, but they are less clear-cut: Only /p/, /s/ and /n/ belong to only one Tukey group, all other consonants belong to more than one. This means that, for instance, /p/ (group ‘a’) does not differ significantly from /t/ (group ‘ab’), which in turn does not differ from /f/ (group ‘bc’) in terms of minimum intensity. Still, /f/ differs from /p/. Group ‘g’ contains all sonorants, so this subdivision is

also visible for intensity. Then again, all sonorants except /n/ belong to group ‘f’, which also the obstruents /z/ and /b/ belong to. Group ‘d’ contains the glides /j/ and /w/, but also the obstruents /v/ and /d/. So although minimum intensity exhibits more groupings, it does not provide more clear distinctions than periodic energy.

Moreover, we can observe an interesting discrepancy between the correlation of the minima’s mean values and the respective sonority indices on the one hand, and their Tukey groupings on the other: Both intensity and periodic energy have similar correlation values (averaged correlation: 0.843 (intensity) vs. 0.851 (periodic energy), cf. Table 3), but periodic energy is much better at creating groups that match the subdivisions of the sonority hierarchies analysed in this study.

Looking at Figure 5, it becomes visible that, although minimum intensity approximately shows the expected trend, the consonants are quite close to each other and do not cover a wide range. This illustrates the unclear Tukey groups, although some of the groups can be perceived visually: First, group ‘a’ contains the voiceless stops, which extend over a similar range in the plot. Other than that, however, no clear distinction between voiceless and voiced obstruents is visible, especially since /s/ tends to pattern more with the voiced ones. Second, the sonorants can be visually grouped together, reflecting Tukey group ‘g’.

Minimum periodic energy (see Figure 6), on the other hand, provides for a much better separation between obstruents and sonorants. Moreover, although the Tukey groups do not subdivide obstruents into voiceless and voiced ones, this separation is supported by the mean values (cf. Table 4) and the plot: The voiceless obstruents consistently reach zero and show no visible variation. This result is expectable because voiceless sounds as such do not have any periodic components. As illustrated above, the only periodic energy measured for voiceless sounds has to be attributed to the surrounding sounds and does not have much effect on the minimum peak. The voiced obstruents, as Figure 6 shows, show more variation: Some speakers produce them with much more minimum periodic energy than voiceless obstruents. However, other speakers exhibit a minimum near zero, which, apparently, makes for a too high intersection between voiceless and voiced obstruents.

The plots illustrate that both measurements produce no perfect correlation with any of the three hierarchies, but the major classes (voiceless obstruents, voiced ob-

struents, sonorants) are ordered in the expected way in both cases, even if not significantly differing from each other.

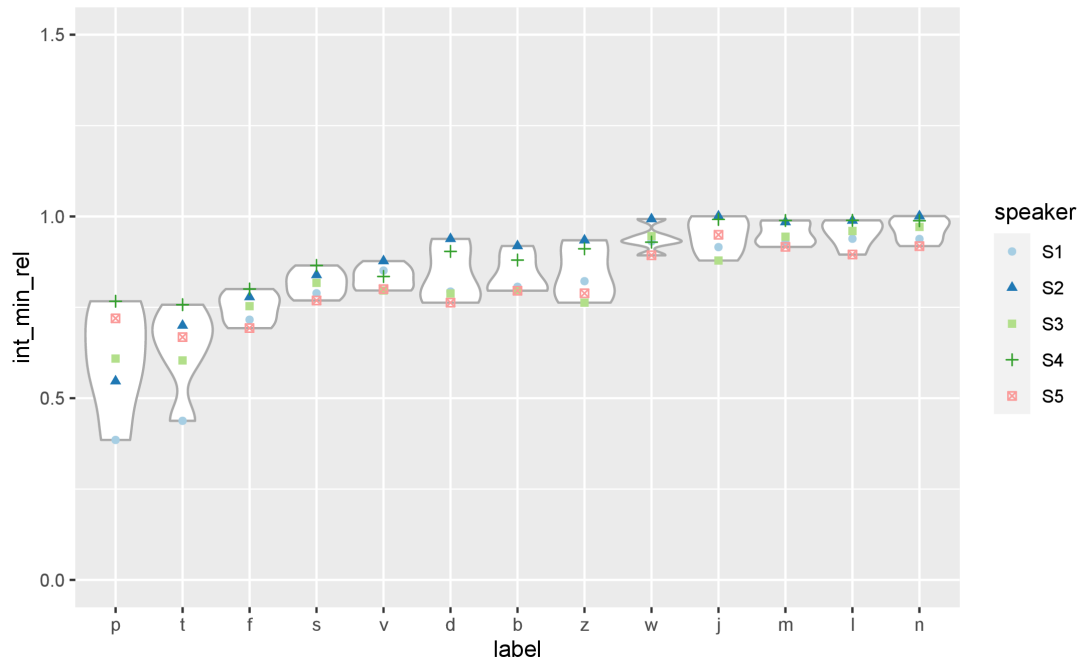


Figure 5: Relative minimum intensity per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

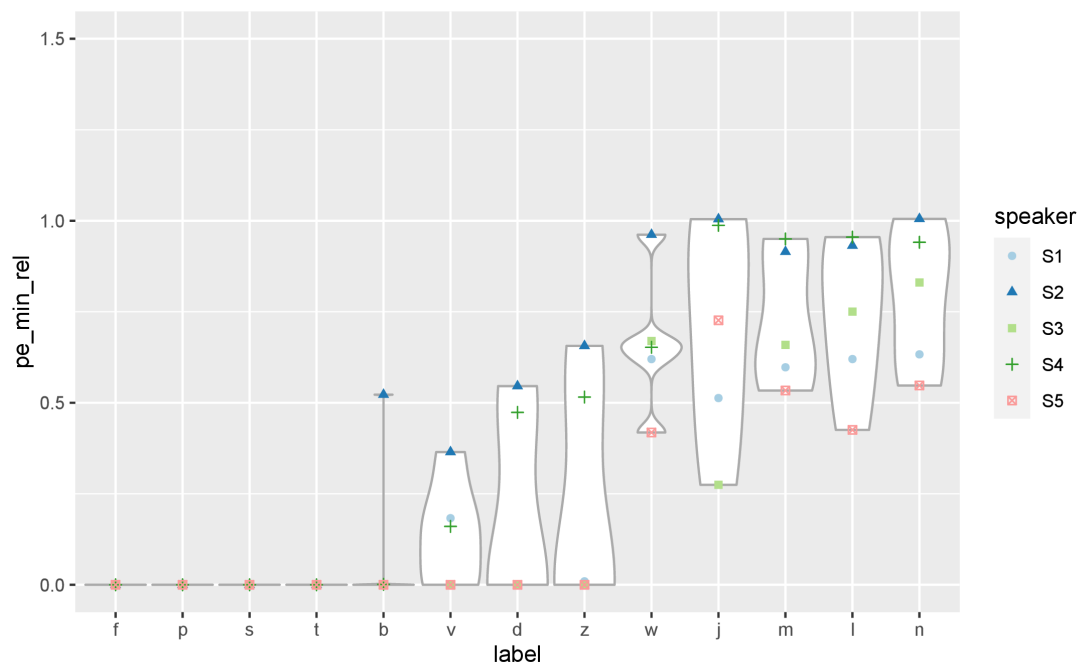


Figure 6: Relative minimum periodic energy per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

3.3.1.2 Maxima

consonant	Maximum intensity	Tukey groups (intensity)	Maximum periodic energy	Tukey groups (periodic energy)
j	1.002	f	1.017	f
w	0.983	cf	0.891	cdf
l	0.985	def	0.894	cdf
m	0.994	ef	0.946	ef
n	0.991	ef	0.944	df
v	0.958	ac	0.720	ac
z	0.942	a	0.667	ab
b	0.980	cf	0.731	ad
d	0.984	cf	0.796	bcde
f	0.944	a	0.569	a
s	0.951	ab	0.559	a
p	0.971	bce	0.730	ad
t	0.962	acd	0.644	ab

Table 5: Relative maximum values of intensity and periodic energy for each consonant with their Tukey groups. The displayed values are averaged across all speakers.

As can be seen in Table 5, neither intensity nor periodic energy maxima let a very clear picture emerge. Starting with intensity, group ‘a’ only contains obstruents (/t/, /s/, /f/, /z/, /v/) – in particular, all fricatives and one stop. Group ‘b’ (/p/, /s/) is rather small and not very informative, as it contains one stop and one fricative, produced at different places of articulation. Group ‘c’ (/t/, /p/, /d/, /b/, /v/, /w/) looks almost like a mirrored version of ‘a’: it contains all stops and only one fricative. However, it does not only contain obstruents, but also one glide. Since glides are usually deemed the most sonorous consonants and, in principle, are regarded as non-syllabic vowels, this finding is clearly not expected. Similar to ‘b’, group ‘d’ (/t/, /l/) is not very useful as it contains one stop and one liquid. In the discourse on sonority, there is no doubt that these two consonants clearly belong to separate groups. Group ‘e’ (/p/, /n/, /m/, /l/) seems a little more promising: it entails the sonorants except the

glides, but, oddly, a stop as well. The final group, ‘f’ (/d/, /b/, /n/, /m/, /l/, /w/, /j/), contains all sonorants and the voiced stops.

As for periodic energy, group ‘f’ encompasses all sonorants. However, this is the only clear match between the maxima Tukey groups and the classes in the sonority hierarchies. Group ‘a’ is almost a perfect counterpart, containing all obstruents except /d/. All other groups are, similar to those of the intensity maxima, not very informative, since no tendency of a match between them and sonority groupings can be observed.

For this measurement, the correlation values roughly reflect the tendency that the Tukey groups show. Periodic energy maxima exhibit higher correlation values than intensity maxima (averaged correlation: 0.713 (intensity) vs. 0.821 (periodic energy), cf. Table 3) and, at the same time, produce a slightly better-fitting division into Tukey groups. However, this difference is still rather small. More salient and important is the fact that, despite the rather high correlation values, both measurements do not show a useful and clear division into Tukey groups. This is reflected in Figures 7 and 8, which show a considerable amount of variation within the consonants, but no clear distinction between them. It can be seen that the values for periodic energy cover a much wider range than intensity. Moreover, for periodic energy, the consonants are not that close to each other. However, although periodic energy exhibits a Tukey group with all sonorants, this grouping is not clearly visible.

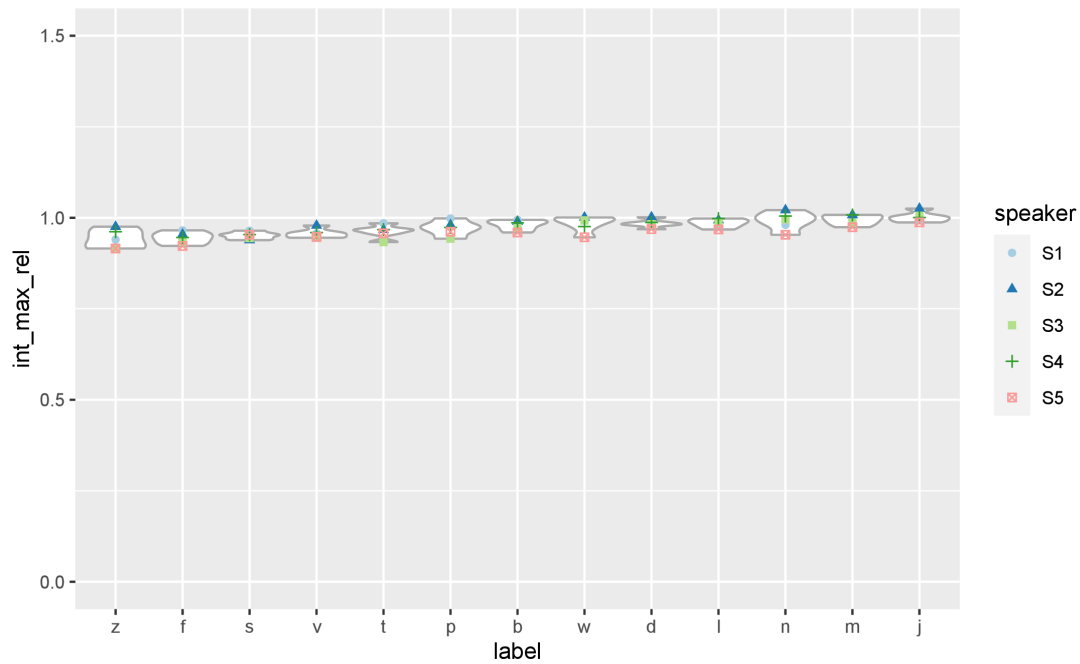


Figure 7: Relative maximum intensity per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

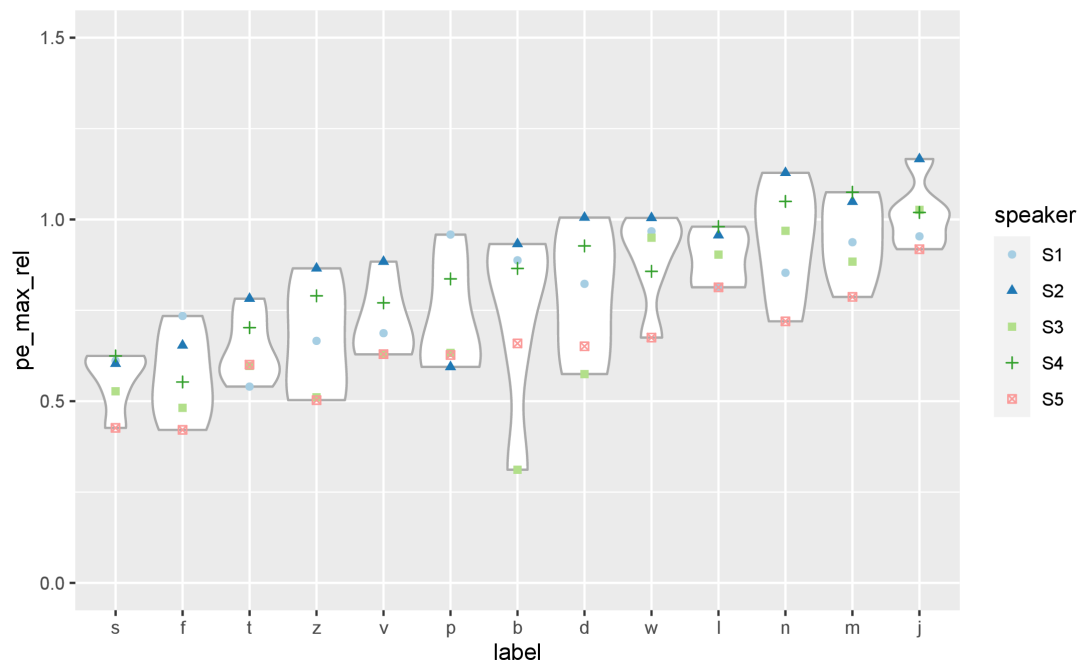


Figure 8: Relative maximum periodic energy per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

3.3.2 Sum

consonant	Summed intensity	Tukey groups (intensity)	Summed periodic energy	Tukey groups (periodic energy)
j	0.695	bc	0.670	c
w	0.537	ab	0.493	bc
l	0.580	ac	0.518	c
m	0.638	ac	0.626	c
n	0.585	ac	0.588	c
v	0.526	ab	0.238	ab
z	0.415	a	0.166	a
b	0.545	ac	0.217	a
d	0.527	ab	0.219	a
f	0.658	ac	0.072	a
s	0.788	c	0.072	a
p	0.695	bc	0.145	a
t	0.734	bc	0.110	a

Table 6: Relative sum values of intensity and periodic energy for each consonant with their Tukey groups. The displayed values are averaged across all speakers.

Moving on to the sums of the variables, Table 6 shows a striking difference between the two measurements: Just as for the minima, periodic energy exhibits a subdivision into obstruents (group ‘a’) and sonorants (group ‘c’). Additionally, group ‘b’ contains /v/ and /w/, which share the place of articulation, but belong to different classes in sonority hierarchies. Again, the mean values suggest a further distinction of voiceless and voiced obstruents, even though this difference is not significant.

Intensity, on the other hand, shows no clear grouping whatsoever: Group ‘a’ contains almost all consonants except for /t/, /p/, /s/ and /j/, and thus cannot be linked to any usual (or, at least, plausible) group. Group ‘b’ (/t/, /p/, /d/, /v/, /w/, /j/) is smaller than ‘a’, but nonetheless puts obstruents and sonorants together by no clear pattern. Finally, group ‘c’ entails all consonants but /d/, /z/, /v/ and /w/. Again, no pattern can be observed.

The difference in mean value correlation (averaged correlation: -0.215 (intensity) vs. 0.866 (periodic energy), cf. Table 3) is reflected by the Tukey groups. As is apparent in Table 6, there is no correlation observable for summed intensity – it is particularly striking that /s/ has the highest mean value. The correlation value for summed intensity already signals that it is possibly no good correlate, which is confirmed by this Tukey test. Figure 9 illustrates this: Neither does the arrangement by mean values produce an order similar to the scales, nor can consonants be clearly distinguished from one another by their measured values.

In Figure 10, on the other hand, it is visible that the periodic energy sum leads to the expected order when looking at the major classes: first, we have voiceless obstruents, then voiced obstruents and, finally, sonorants. There is a visible separation into obstruents and sonorants. Furthermore, one can almost separate the obstruents into voiceless and voiced ones, since most voiceless obstruents have smaller mean values and less variation. However, /p/ exhibits much more variation than the other voiceless obstruents, and the overlap between both classes is too high for a significant difference. The periodic energy sums go down to a minimum near zero, whereas with approx. 0.25, this minimum is much higher for intensity.

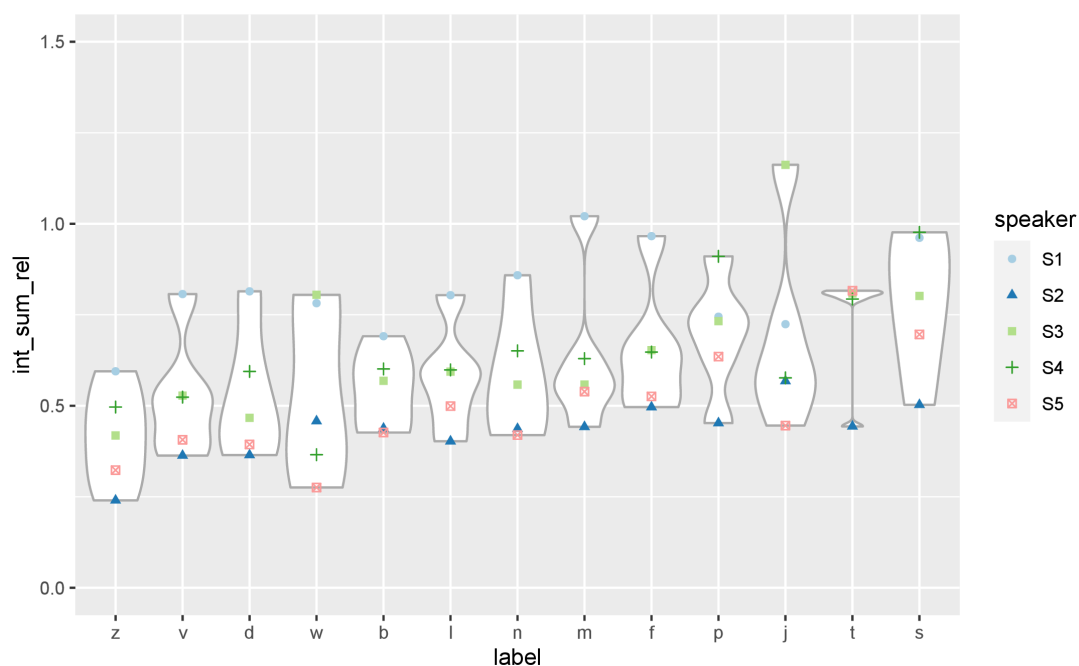


Figure 9: Relative summed intensity per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

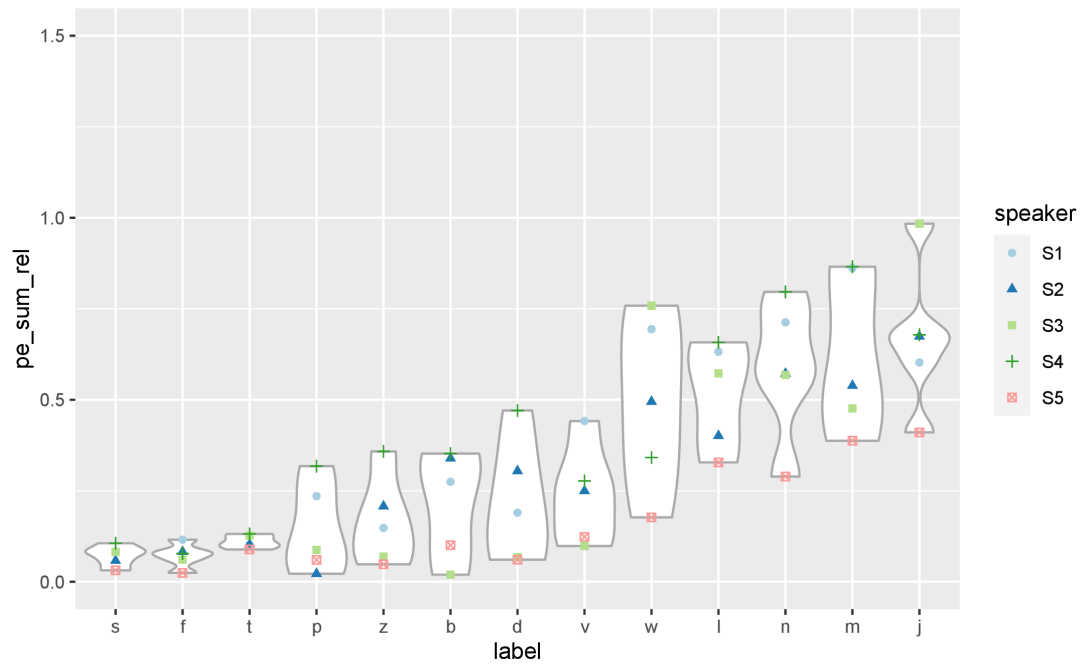


Figure 10: Relative summed periodic energy per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

3.3.3 Average

consonant	Averaged intensity	Tukey groups (intensity)	Averaged periodic energy	Tukey groups (periodic energy)
j	0.998	c	0.995	c
w	0.982	c	0.885	c
l	0.981	c	0.887	c
m	0.999	c	1.011	c
n	1.000	c	1.024	c
v	0.904	b	0.413	b
z	0.901	b	0.402	ab
b	0.915	b	0.381	ab
d	0.913	b	0.410	b
f	0.830	a	0.092	ab
s	0.875	ab	0.081	a
p	0.839	a	0.158	ab
t	0.828	a	0.131	ab

Table 7: Relative average values of intensity and periodic energy for each consonant with their Tukey groups. The displayed values are averaged across all speakers.

The last measuring technique is averaging, for which the results will be discussed in this subsection. Table 7 shows the results for intensity and periodic energy. Just as with the results for summing, both intensity and periodic energy produce only three different groups. Periodic energy, again, can divide the consonants into sonorants and obstruents, encompassing all sonorants in group ‘c’. The obstruents, however, show a less clear picture: Group ‘a’ contains all obstruents except for /d/ and /v/, whereas in group ‘b’, all obstruents except /s/ are included. Although the mean values suggest a subdivision into voiceless and voiced obstruents, the Tukey test does not confirm this, indicating that there is a great deal of variation within the single values.

With the other measurements, periodic energy tends to match sonority hierarchies better than intensity. Interestingly, for averaging, this is not the case: First, intensity performs the same division into sonorants and obstruents, as group ‘c’ exclusively contains sonorants. Second, a meaningful subdivision of the obstruents can be observed, too: Group ‘a’ entails all voiceless obstruents, group ‘b’ all voiced obstruents

and, in addition, /s/. Disregarding this potential artefact, average intensity comes closest – out of all measurements – to hierarchy (3) by Albert and Nicenboim (2020), which otherwise only separates glides from the other sonorants.

Since the average values are based on the same calculation as summing and, as the only additional step, normalise this by the consonants' duration, a comparison between these is useful. For periodic energy, summing produces a clearer result than averaging, although the main subdivision (sonorants vs. obstruents) becomes apparent in both cases and both have a similar correlation (averaged correlation: 0.866 (sum) vs. 0.836 (average), cf. Table 3). Regarding intensity, there is a great discrepancy between summing and averaging: Summing does not lead to usable results at all, as no correlation with any of the three examined sonority hierarchies is observable and no meaningful Tukey group can be established. Averaging, on the other hand, produces very clear groups, even the clearest of all investigated measurements. However, it can be concluded that periodic energy is more robust with regard to duration, because useful results are obtained in both cases and not only in the average.

Similar to summing, the correlation values for averaging (averaged correlation: 0.847 (intensity) vs. 0.836 (periodic energy), cf. Table 3) are a proper reflection of the goodness of fit for the Tukey groups. However, although intensity has a slightly better correlation and slightly better-fitting Tukey groups, it is important to note that the values do not cover a wide range and the effects are rather small (see Figure 11) compared to periodic energy (see Figure 12).

Although the values are very close to each other, Figure 11 illustrates the separation of the consonants by their average intensity values into voiceless obstruents, voiced obstruents and sonorants. Nevertheless, /s/ is not clearly separable from either voiceless or voiced obstruents, as indicated by the ambiguous Tukey group assignment 'ab'.

Like minimum and summed periodic energy, the plot for average periodic energy indicates a difference between voiceless and voiced obstruents. However, the great variation within the voiced obstruents produces an overlap with the voiceless ones which is too large for a significant difference. We can see that, similar to the periodic energy sum, /p/ still shows more variation than the other voiceless obstruents, but in relation to the voiced ones, its variation is smaller.

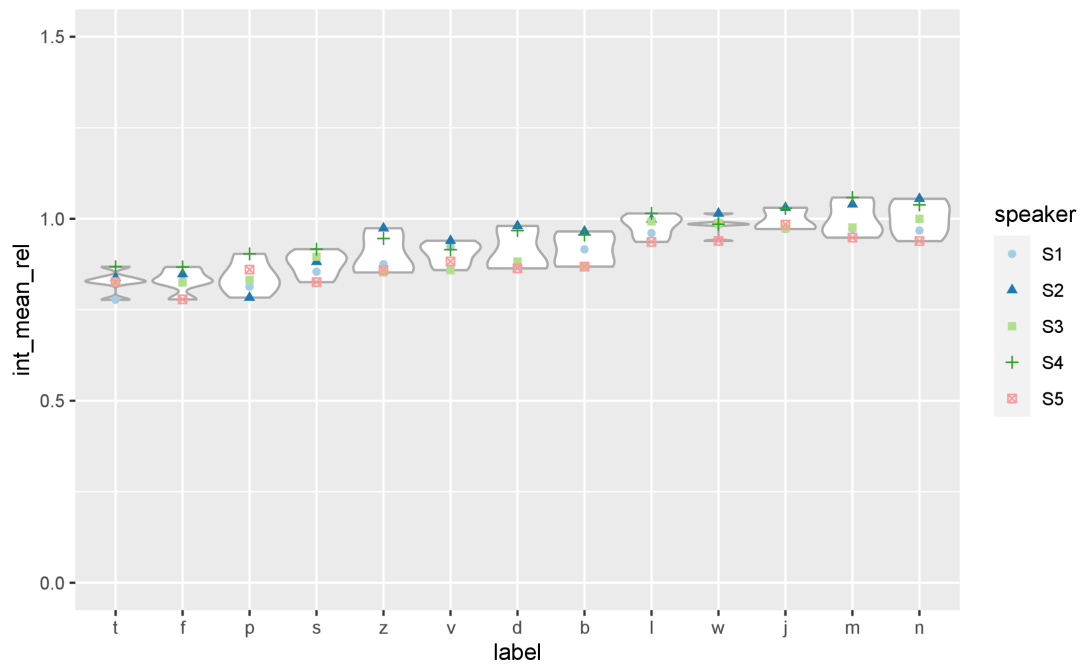


Figure 11: Relative average intensity per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

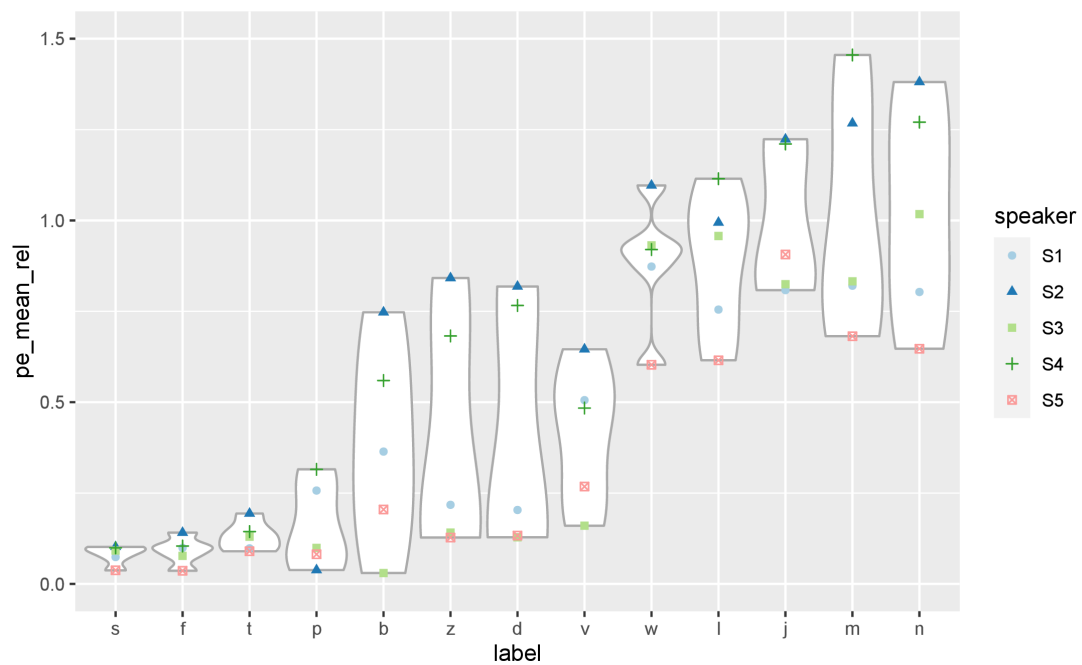


Figure 12: Relative average periodic energy per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

3.4 Discussion

This study provides a large amount of results which lead to interesting findings. In the following subsections, I will discuss the results for summing (subsection 3.4.1), peaks (subsection 3.4.2) and averaging (subsection 3.4.3). Finally, I will present some general remarks on this study (subsection 3.4.4).

3.4.1 The role of duration: Summing makes a difference

Summing produces the highest discrepancy between intensity and periodic energy: The periodic energy sum has very good correlation values, even the best among all measures (averaged across all hierarchies), and produces useful Tukey groups. Intensity, on the other hand, produces the worst values of all, with a correlation near zero or low negative correlations. Aside from that, no useful Tukey groups can be established. Thus, summed intensity does not produce any sonority-relevant grouping pattern, whereas summed periodic energy turns out to exhibit a general pattern of sonority-relevant grouping.

Why is this the case? First, one has to keep in mind how the investigated variables relate to each other: Intensity refers to the power of the acoustic signal. Periodic energy can be seen as a subpart of general intensity, as it refers to the intensity of the signal's periodic components. Since only voiced consonants exhibit periodic components, voiceless consonants are bound to have a low periodic energy sum – their only periodic components have to be attributed to surrounding voiced segments. This is in line with the fact that voiceless sounds are considered least sonorous in all examined sonority hierarchies. In contrast, intensity is not necessarily low for voiceless sounds, as it measures the whole power of the signal, including aperiodic components such as friction. This becomes apparent in the analysis, where /s/ has, on average, the highest intensity sum, and all other voiceless sounds tend to rank high, too. Moreover, even though voiced obstruents have periodic components, the rather high constriction in the vocal tract dampens them. This, again, fits most sonority hierarchies that are part of this study, as voiced obstruents are often considered more sonorous than voiceless ones, and always considered to be less sonorous than sonorants.

For these reasons, summed periodic energy is generally expected to produce good results. Why, on the other hand, does summed intensity produce so bad results, espe-

cially given the fact that intensity produces rather good results in other measurements? Particularly, average intensity leads to good correlation and grouping results. Importantly, the average measurement is related very closely to summing: A consonant's sum (of intensity or periodic energy) is divided by its duration to neutralise its accumulative contribution. As is visible in Figure 13, the consonants are of different durations. For example, /s/ has a very long duration, potentially leading to a greater sum value than a short consonant, e.g., /z/. Recalling that /s/ has a high intensity sum, a division by its long duration leads to a lower average in comparison to the other consonants. In contrast, its periodic energy sum is already low (near zero), so it remains low, regardless of whether duration is accumulated (sum) or neutralised (average). This applies to all voiceless sounds: They all have a long duration and a high intensity sum, but their periodic energy sum is low.

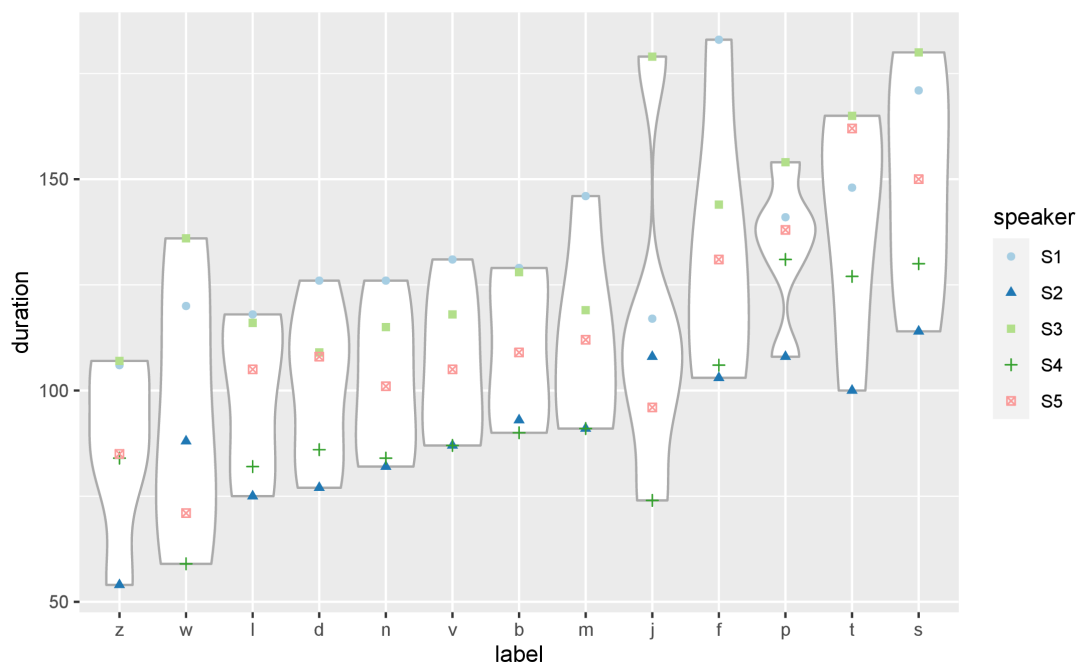


Figure 13: Duration (ms) per speech sound (x axis) and speaker (symbols). The sounds are arranged according to their mean value.

There is good evidence to support the inclusion of duration as a contributing factor to sonority. For instance, Price (1980) provides data from a syllable-count perception task. In acoustical recordings, the duration of a sonorant (/l/ or /r/) was manipulated and participants were asked whether they heard, e.g., the monosyllabic

word ‘plight’ /plʌɪt/ or the disyllabic word ‘polite’ /pəˈlaɪt/. The longer the duration of the sonorant, the more the word is perceived as disyllabic, and, as a consequence, the more sonorous does the consonant appear.

This shows that duration plays a role in the sonority status of voiced consonants. The periodic energy sum takes this into account, with longer duration leading to higher sonority. Long voiceless sounds, on the other hand, still have a periodic energy sum near zero and presumably do not substantially increase in sonority. For example, a lengthened /s/ in ‘scum’ /skʌm/ would probably not make people perceive the word ‘succumb’ /səˈkʌm/: /s/ has no periodic energy and should therefore not lead to the perception of a vowel.

3.4.2 The role of segmentation: Why min is better than max

The peak measurements all produce rather good correlation values (between 0.71 and 0.85), but only minimum periodic energy produces a clear subdivision of the consonants, namely into obstruents and sonorants. This subdivision is somewhat noticeable for minimum intensity and maximum periodic energy, too, but much less clear-cut. Maximum intensity does not produce any useful Tukey groups at all.

The fact that minima score better than maxima can be explained as follows: Mostly, consonants exhibit lower values of intensity and periodic energy than surrounding vowels. Hence, consonantal minima of intensity and periodic energy are likely to occur in the middle of their segment. This makes the minima very robust to segmentation: Slight changes in position of the segmental boundaries should have no effect on the minimum, which should still be found (approximately) in the consonant’s centre. Maxima, on the other hand, often occur right at the segmental boundaries – with the setup of this study, because every measured consonant is surrounded by vowels. Thus, the maxima might not be attributed to the consonant as such, but to the preceding or following vowel. In such a case, segmentation has a direct impact on the consonant’s maximum value and is likely to skew the results. This is especially true for periodic energy of voiceless sounds: They do not exhibit any periodic energy themselves, so their maximum periodic energy can only originate from the surrounding vowels.

This issue can also explain the fact that absolute maximum intensity scores far worse than its relative counterpart: Variation in the intensity of the following vowel

very likely leads to variation in the consonant's maximum. When normalising in relation to the following vowel's maximum intensity, this variation is decreased and the results appear less skewed.

The Tukey analysis of the peak measurements shows that it is not sufficient to only compare correlation values. Furthermore, this study shows that a minimum measurement that averages closure and burst of aspirated consonants, as Parker (2008) proposed, does not have any benefit on the correlation values. In summary, minimum values can, especially with periodic energy, form a rather good correlate for sonority. It should be noted, however, that peak values (minimum and maximum) only consider one point in time. In the case of laboratory experiments, minima appear to be a rather robust cue. Though, they might be sensitive to, e.g., background noise in everyday recordings, or signal distortions. Summing and averaging, in contrast, take the whole consonant's duration into account and are thus more robust against that.

3.4.3 The mean between: The robustness of the average measure

In contrast to summing and peak measurements, the average measure turns out to produce good results in every case: both intensity and periodic energy, both the absolute and relative version. The reasons as to why the intensity average performs much better than summed intensity have been discussed above and are summarised again here: Especially voiceless sounds, i.e. sounds with the lowest sonority values, tend to have a long duration that can play a role when summing is applied on a general acoustic dimension such as intensity. Averaging, in contrast to summing, neutralises the duration component, which results in lower intensity measurements for these sounds. Their periodic energy, on the other hand, is already near zero, so periodic energy measures remain low for voiceless obstruents even when the values accumulate for the duration of the consonant. Moreover, the average intensity values cover a smaller range than average periodic energy. This reflects the fact that the intensity sum values are very close to each other and the normalisation by duration does produce significant differences, but still, rather small ones.

As stated above, there is evidence that duration is relevant with respect to sonority – especially for voiced sounds. Following this assumption, the periodic energy sum appears as the most suitable measurement: The sum increases with a longer duration,

and periodic energy leads to the desirable difference in the accumulative effect of summing, which is strong for sonorants and weak for obstruents, especially voiceless. An average, in contrast, does not change substantially with a change in duration. Thus, this type of measurement would not account for the fact that, e.g., a long sonorant could be perceived as more sonorous. Nevertheless, it still provides a good approximation, as demonstrated in this study. Especially in cases where, for instance, only intensity values can be examined, or the contribution of duration is not certain, measuring the average is a good approach for a sonority correlate. In addition, it is not as sensitive to minor signal distortions as peaks can be. Although averaging might not be the best approach in every case, the results of this study suggest that it is a very robust one.

3.4.4 General remarks

Across the board, no measurement subdivides the investigated consonants as fine-grained as any of the considered sonority hierarchies. The most frequent subdivision is into obstruents and sonorants. One measurement additionally produces a significant distinction of voiceless and voiced obstruents (with one ambiguously assigned consonant), and some more measures show a tendency towards this when averaging across all speakers. A follow-up study with more participants might produce a clearer picture. Moreover, it is still important to note that most measurements result in rather high correlation values and thus encompass at least some aspect of sonority.

Regarding the sonorants, this study does not provide evidence for a further subdivision based on the conducted measurements. Especially, /w/ often exhibits the lowest values of all sonorants, but being a glide, it has the highest sonority values in all of the three investigated sonority hierarchies. It has to be noted, though, that the values show a considerable amount of variation. Again, a study with more participants and more stimuli could produce a different picture.

Generally speaking, this study is of limited statistical power, due to the low number of participants. A larger amount of stimuli – and perhaps more natural and diverse ones – would also be beneficial for a following study's explanatory power. However, when changing the stimuli from pseudowords to natural words, possible effects of lexical frequency should be considered.

4 Conclusion

This exploratory study contributes to the ongoing debate on the nature of sonority and the question whether it can be measured physically. To approach this problem, I conducted several acoustic measurements on recorded consonants and calculated correlations with three different prototypical sonority hierarchies. Following, I analysed the different types of measurements more closely in order to find possible groups among the consonants.

The results show that most measurements lead to rather high correlation values. Moreover, the consonants often subdivide into obstruents and sonorants, and a tendency to divide obstruents into voiceless and voiced ones is visible. This shows that sonority can, at least in part, be measured physically.

None of the measurements produces a perfect correlation with any of the considered sonority hierarchies. However, a perfect correlation might not even be realistic, since speech sounds are not bound to have fixed sonority values. For instance, there is evidence for effects of segmental duration on sonority: Voiced consonants tend to be perceived as more sonorous when lengthened. This contradicts the idea of fixed sonority values, which all symbolic sonority hierarchies are based on.

Nevertheless, the measurement of summed periodic energy actually takes the effect of duration into account: Voiced sounds, especially sonorants, have a substantially higher periodic energy sum with a longer duration. Thus, this measurement supports the idea that speech sounds do not have fixed sonority values and, probably, no perfect correlate is expectable. This reflects the general problem of examining continuous natural language by means of discrete symbols.

On the one hand, summed periodic energy supports the idea that a perfect correlation could be impossible. On the other hand, of all measurements in this study, it still comes closest to being perfect. This additionally corroborates the assumption that periodic energy is essential to sonority.

Looking at the average measurement, both examined variables perform well. If periodic energy cannot be obtained, average intensity seems to be a good approximation to sonority – in contrast to summed intensity, which is not useful at all. Of the peak values, especially minimum periodic energy shows good results. Minimum in-

tensity shows a good correlation as well, but less clear groups of consonants. Finally, at least for consonants, maximum peaks are no reliable cue to sonority.

As already stated, this is an exploratory pilot study with few participants and therefore limited statistical power. Still, it provides interesting findings that can be the basis of further examination. Future studies with more participants could analyse more speech sounds in varying prosodic conditions. Moreover, a broad bandwidth of languages would be beneficial for a cross-linguistic judgment on sonority.

This study did not find significant differences within the group of sonorants, although many proposed sonority hierarchies further subdivide this group. It should be interesting to see whether follow-up studies confirm any subdivision. The results can then be taken as an indication of how fine-grained a sonority hierarchy should be designed. This study suggests a rather coarse setup of such hierarchies, particularly considering the idea that speech sounds are not likely to have fixed sonority values.

References

- Albert, Aviad, Francesco Cangemi & Martine Grice (2018). Using periodic energy to enrich acoustic representations of pitch in speech: A demonstration. *Proceedings of Speech Prosody (9)*, Poznań, 804–808.
- Albert, Aviad & Bruno Nicenboim (2020). *Take a NAP: A new model of sonority using periodic energy and the Nucleus Attraction Principle*. Manuscript submitted for publication.
- Albert, Aviad, Francesco Cangemi & Martine Grice (2020). Periogram Projekt: Workflows for periodic energy extraction and usage. OSF. Obtained on 2020-03-11. doi:10.17605/OSF.IO/28EA5.
- Bell-Berti, Fredericka & Katherine S. Harris (1981). A temporal model of speech production. *Phonetica* 38(1–3): 9–20.
- Boersma, Paul & David Weenink (2020). Praat: Doing phonetics by computer. Version 6.1.13. Retrieved from <http://www.praat.org/>
- Browman, Catherine P. & Louis Goldstein (1992). Articulatory phonology: An overview. *Phonetica* 49(3–4), 155–180.
- Clements, George Nickerson (1990). The role of the sonority cycle in core syllabification. In Kingston, John & Mary E. Beckman (eds.), *Papers in laboratory phonology 1: Between the grammar and physics of speech*. Cambridge: Cambridge University Press, 283–333.
- Gordon, Matthew (2004). Syllable weight. In Hayes, Bruce (ed.), *Phonetically based phonology*. Cambridge: Cambridge University Press, 277–312.
- Gordon, Matthew, Edita Ghushchyan, Bradley McDonnell, Daisy Rosenblum & Patricia A. Shaw (2012). Sonority and central vowels: A cross-linguistic phonetic study. In Parker, Steve (ed.), *The Sonority Controversy*. Berlin: De Gruyter Mouton, 219–256.
- Heselwood, Barry (1998). An unusual kind of sonority and its implications for phonetic theory. *Leeds Working Papers in Linguistics & Phonetics* 6, 68–80.
- ITU-R Recommendation BS.1770-4 (2015). *Algorithms to measure audio programme loudness and true-peak audio level*. Geneva: ITU. BS Series: Broadcasting service (sound).

- Jany, Carmen, Matthew Gordon, Carlos M. Nash & Nobutaka Takara (2007). How universal is the sonority hierarchy?: A cross-linguistic acoustic study. *Proceedings of the 16th ICPHS*, Saarbrücken, 1401–1404.
- Kawai, Goh & Jan van Santen (2002). Automatic detection of syllabic nuclei using acoustic measures. *Proceedings of 2002 IEEE Workshop on Speech Synthesis*, Santa Monica, 39–42.
- Kawasaki, Haruko (1982). *An acoustical basis for universal constraints on sound sequences*. PhD thesis. UC Berkeley: Department of Linguistics.
- Komatsu, Masahiko, Shinichi Tokuma, Won Tokuma & Takayuki Arai (2002). Multi-dimensional analysis of sonority: Perception, acoustics, and phonology. *Interspeech 2002: 7th International Conference on Spoken Language Processing*, Denver, Colorado, 2293–2296.
- Ladefoged, Peter (1997). Linguistic phonetic descriptions. In Hardcastle, William J. & John Laver (eds.), *The handbook of phonetic sciences*. Oxford; Cambridge, MA: Blackwell, 589–618.
- Llanos, Fernando, Joshua M. Alexander & Christian E. Stilp (2015). Shannon entropy predicts the sonority status of natural classes in English. Poster presented at the *169th Meeting of the Acoustical Society of America*, Pittsburgh, Pennsylvania.
- Nakajima, Yoshitaka, Kazuo Ueda, Shota Fujimaru, Hiroto Motomura & Yuki Oh-saka (2012). Acoustic correlate of phonological sonority in British English. *Proceedings of Fechner Day 28*, 56–61.
- Ohala, John J. (1992). Alternatives to the sonority hierarchy for explaining segmental sequential constraints. *Papers from the parasession on the syllable*, Chicago, 319–338.
- Parker, Steve (2002). *Quantifying the sonority hierarchy*. PhD thesis. Amherst: University of Massachusetts.
- Parker, Steve (2008). Sound level protrusions as physical correlates of sonority. *Journal of phonetics* 36(1), 55–90.
- Parker, Steve (2011). Sonority. In van Oostendorp, Marc, Colin J. Ewen, Elizabeth Hume & Keren Rice (eds.) *The Blackwell companion to phonology 2*. West Sussex, UK: Wiley-Blackwell, 1160–1184.

- Parker, S. (2017). Sounding out sonority. *Language & Linguistics Compass* 11(9), 1–197.
- Patha, Sreedhar, Yegnanarayana Bayya, and Suryakanth V. Gangashetty (2016). Syllable nucleus and boundary detection in noisy conditions. *Proceedings of Speech Prosody (8)*, Boston, 360–364.
- Price, Patti J. (1980). Sonority and syllabicity: Acoustic correlates of perception. *Phonetica* 37(5–6), 327–343.
- R Core Team (2020). R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. Retrieved from <https://www.R-project.org/>
- Selkirk, Elisabeth (1984). On the major class features and syllable theory. In Aronoff, Mark, Richard T. Oehrle & Morris Halle (eds.), *Language sound and structure: Studies in phonology presented to Morris Halle by his teacher and students* (pp.). Cambridge, MA: The MIT Press, 107–136.
- Sievers, Eduard (1893/1901). *Grundzüge der Phonetik zur Einführung in das Studium der Lautlehre der indogermanischen Sprachen*. Leipzig: Breitkopf & Härtel.
- Steriade, Donca (1982). *Greek prosodies and the nature of syllabification*. PhD thesis. Massachusetts: MIT.

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Hiermit versichere ich an Eides statt, dass ich diese Bachelorarbeit selbstständig verfasst und ohne die Benutzung anderer als der angegebenen Hilfsmittel angefertigt habe.

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