A CROSS-LANGUAGE ACOUSTIC SPACE
FOR VOCALIC PHONATION DISTINCTIONS

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REVISED DRAFT, NOV 2021
ABSTRACT

Many languages use multiple phonation types for phonemic or allophonic distinctions. This study examines the acoustic structure of the phonetic space for vowel phonations across languages, focusing on the acoustic phonetic space for languages with non-modal phonation on vowels (rather than on consonants, or as coarticulation from consonant contrasts). Our sample of 11 languages, from five language families, includes languages with contrastive phonation types on vowels, allophonic non-modal phonation associated with particular tones, and English as a single-category case. Together these 11 languages provide 29 instances from among the following categories: Modal, Breathy, Creaky, Lax, Tense, Harsh, and/or Pharyngealized. Acoustic measures of vowel samples were made using VoiceSauce, and Multi-Dimensional Scaling was used to obtain from these measures a low-dimensional acoustic space within which the 29 categories can be compared. This space is largely two-dimensional; English lies in the middle of the space. !Xóõ, with five contrastive phonation categories, uses the largest space, and requires a third dimension to fully distinguish all five. The first dimension varies from modal to non-modal phonations, and is important for distinguishing allophonic creaky voice from modal.

The second dimension is like a traditional Breathy-to-Creaky continuum, and seems to be basic for phonation contrasts: all languages with a contrast use it, and languages with just two categories make them on this dimension. The dimensions can be related back to the acoustic measures that structure them, indicating which measures are most important across languages. From our MDS results, and a separate classification and regression tree analysis of all individual datapoints, we can recommend H1*–H2*, Subharmonic to Harmonic Ratio, Harmonic to Noise Ratio 0–500 Hz, and optionally H1*–A1*, Strength of Excitation, and Energy, as a small set of the most informative parameters that could be included in future studies.*

*Acknowledgements to follow in non-anonymous version

Keywords: Phonetic typology, phonation types, voice quality, phonetic categories, breathy voice, creaky voice
1. **Introduction.** Phonation is the production of sound in the larynx. Often this term is used in a narrow sense to refer only to the production of voicing, i.e. vibration of the vocal folds inside the larynx, but more broadly, it refers to the production of aperiodic noise as well. Human voices can vary in the rate (frequency) of vibration of the vocal folds, which we hear as changes in voice pitch; but they can also vary in the spatial pattern of the folds’ vibratory movements. We hear these kinds of variations as differences in phonatory quality. Each individual speaker can manipulate their vocal folds during speech so as to produce a range of pitches and qualities; we do this for both linguistic purposes (e.g. prosody, coarticulation) and paralinguistic purposes (e.g. emotion) (Laver 1980, Garellek 2012, Yanushevskaya, Gobl, & Ni Chasaide 2018, Davidson 2021, and references therein).

Phonatory quality in turn contributes to overall voice quality, which differentiates individuals’ voices and accents. It is also the case that languages may differ in the pitch ranges and voice qualities that they typically use (Esling et al. 2019). For example, studies have shown that bilinguals can use two different voice qualities when speaking their two different languages (Bruyninckx 1994, Engelbert 2014), and that pairs of languages can have measurably perceptually different phonatory qualities (Yiu et al. 2008). That is, voice quality is one of the many ways in which languages can sound different from one another; put another way, having a native accent in a language involves using an appropriate range of voice qualities for that language.

At the same time, many languages use multiple phonation types (or phonation categories) for phonemic contrasts. In such languages, each speaker must produce a set of distinctive laryngeal phonatory qualities in order to distinguish word meanings. Phonation variation then cannot depend only on prosody or emotion, and it is not entirely up to the speaker how much variation to show. Instead, each speaker has to consistently employ a range of phonatory qualities, and do so in a way similar to other speakers of their language. What is the space of possibilities available to languages for such phonation contrasts? What laryngeal articulations, what acoustic dimensions, what auditory qualities are accessible to populations of speakers and give reliably contrasting categories? We envision a multi-dimensional phonetic space for phonation, within which different languages locate their sets of phonation categories.
The main question we address in this paper is the acoustic structure of this phonetic space for phonation types across languages. There are two prerequisites to this research. First, we need to know about the languages of the world and their linguistically-relevant phonation types, from which we can select a (small) sample for our study. Second, we need appropriate tools for semi-automatic phonetic analysis at this scale, since even a small sample of languages will involve thousands of individual speech sound tokens. Such analysis is much more practical in the acoustic domain, and therefore we direct our research to questions about the acoustic phonetic space for phonations. And, because acoustic phonation analysis is more reliable when limited to vowel sounds, we limit our research to the acoustic phonetic space for phonation in vowels, and thus to languages with non-modal phonation on vowels (rather than on consonants, or as coarticulation from adjacent consonant contrasts). Our sample includes languages with contrastive phonation types on vowels, as well as those with allophonic non-modal phonation associated with particular tones.

1.1. Vowel phonation in languages. Ladefoged and Maddieson, in their 1996 compendium of segmental contrasts in the world’s languages, discuss phonation types as a minor feature of vowel contrasts. Their phonation types are the same ones as they describe for consonants: breathy voice, slack or lax voice, modal voice, stiff or tense voice, and creaky voice. These are said to form a continuum in terms of the airflow through the glottis: at one extreme of the continuum is voicelessness with maximum glottal airflow, and at the other extreme is voicelessness with zero glottal airflow, e.g. a glottal stop [ʔ]. The voicing categories fall between these voiceless extremes: breathy and lax voice with greater airflow than modal voice, and creaky and tense voice with less. Ladefoged (1971) also describes a subset of these categories as forming a continuum in terms of glottal opening (specifically, distance between the arytenoid cartilages at the rear of the glottis): voiceless aspiration – breathy – modal – creaky – glottal stop. Lax and tense voice, in contrast, are distinguished by tension in the vocalis muscle.

In a different tradition, Laver and others distinguish many more phonation types, based on the possibilities of the human larynx rather than attested linguistic contrasts (e.g. Laver 1980, Esling & Harris 2005, Gobl & Ni Chasaide 2012, Esling et al. 2019). For example, two kinds of phonation with breathier voice qualities, i.e. those with greater glottal opening and higher
airflow, can be distinguished: whispery voice vs. breathy voice (narrowly defined) (Moisik et al. 2021), where whispery voice has more spread vocal folds but greater epilaryngeal constriction. Here, we will begin with a single phonation category for “breathy voice”, but can use our cross-language analyses to explore whether finer distinctions can be made. Sub-types of creaky voice qualities likewise are possible, as reviewed by Keating et al. (2015). Again, the linguistic relevance of the sub-types is not understood. Therefore, again we begin with a single category of “creaky voice”, but can examine our own data for evidence of language differences in this regard.

An additional phonation type described separately by Ladefoged & Maddieson (1996) is strident (or harsh/epiglottalized) phonation, e.g. in languages within the “Khoisan” group. In addition to changes in vocal fold vibration, characterized by Esling as whispery voice due to vocal fold spreading, this phonation is also said to involve epilaryngeal constriction, aryepiglottic trilling, pharyngeal narrowing, and tongue retraction (Traill 1986, Hess 1998, Esling 2005, Edmondson & Esling 2006, Miller 2007, Moisik et al. 2021).

In their survey paper on phonation contrasts (for both vowels and consonants) across languages, Gordon & Ladefoged (2001) use the continuum of glottal opening to describe and classify phonation types in a wide variety of languages, and suggest measures of the glottal source and filter that can be used to quantify differences across phonation categories. There have since been many instrumental studies of vowel phonation contrasts in individual languages. We now have basic descriptions of such contrasts in dozens of languages, providing a foundation for cross-language comparisons. Figure 1 displays a sample of 61 languages that have phonation contrasts on vowels, including all of the languages with contrasts in the present study. (The list of languages shown here, with further information and references, is given in the Appendix.) This sample is not exhaustive, since we do not include all known examples from within a given sub-family; for example, we include only five of the many Zapotec languages with contrastive creaky voice on vowels.
Figure 1. Geographical distribution of a sample of 60 languages with contrastive non-modal phonation on vowels. Each circle represents a language (some circles cannot be distinguished due to overlap). Blue circles are languages that have breathy (or lax) vowels; yellow circles are languages that have creaky (or tense) vowels; green circles are languages with both breathy and creaky vowels. The figure includes the languages with phonation contrasts presented in this paper (whose names appear in the map), as well as others from sources such as Gordon & Ladefoged (2001) and the PHOIBLE 2.0 online repository (Moran & McCloy 2019).

Languages with non-modal phonation on vowels come from a wide variety of families, but languages from certain families – Otomanguean, Nilotic, Kx’a, and Tuu – regularly exhibit vowel phonation contrasts. Of these, Otomanguean languages (especially those in the Zapotec and Mixtec branches) tend to contrast creaky vs. modal vowels, though some Valley Zapotec languages have a three-way creaky vs. modal vs. breathy distinction in vowels (Munro & Lopez 1999, Esposito 2010b, Ariza García 2018). Nilotic languages tend to contrast breathy vs. modal vowels instead, though some languages in the family have breathy and harsh vowels.
Languages in the Kx’a (e.g. Ju’hoansi) and Tuu (e.g. !Xóõ) families often have breathy and creaky vowels, which further contrast with pharyngealized and/or harsh vowels (Traill 1985, Miller 2007, Garellek 2020).

The high occurrence of vowel phonation contrasts among certain Sino-Tibetan, Hmong-Mien, and Austroasiatic languages is largely geographically restricted to the Southeast Asian Sprachbund, where languages with register and tense–lax contrasts abound (Maddieson & Ladefoged 1985, Zhu 2012, Brunelle & Kirby 2016, Garellek et al. 2021); these kinds of contrasts are defined below.

Outside the aforementioned groups, isolated cases of languages with vowel phonation contrasts can be found across the world, and tend to contrast creaky vs. modal vowels; e.g., Udihe (Tungusic, Russia). However, Gujarati (Indo-European, India) contrasts breathy vs. modal vowels, and has no contrastive creak.

Finally, languages with phonation contrasts on vowels tend to also have tonal contrasts. In our sample, only 25% of languages (mostly Austroasiatic) lack lexical tone; see the list in the Appendix. Many of these have mixed tone–phonation or register systems, where contrastive categories are distinguished by phonation as well as pitch, vowel quality, and duration. In these languages, the non-modal phonation can serve as a primary or secondary cue to a particular register (Brunelle & Finkeldey 2011, Brunelle 2012), as is argued for (White) Hmong (Hmong-Mien; Southeast Asia) and (Black) Miao (Hmong-Mien; China) (Garellek et al. 2013, Kuang 2013). For instance, in Takhian Thong Chong two of the four registers (breathy and breathy-tense) have similar pitch contours, while the other two registers (modal and tense) are fairly well-separated from these and from each other in pitch (DiCanio 2009). See also Esposito & Khan (2020) for discussion of mixed tone-phonation contrasts; other examples of phonation contrasts can be found in Esposito & Khan (2020) and Garellek et al. (2021).

In the present study, we draw from this sample a small but relatively diverse set, to explore the acoustic differences among phonation categories.

1.2. ACOUSTIC TOOLS. Since at least Klatt & Klatt (1990; see also Baken & Orlikoff 2000, Ch. 7) a variety of acoustic measures have been identified that reflect aspects of voice production and voice quality variation. Gordon & Ladefoged (2001) list measures of periodicity, energy, spectral
tilt, pitch, and duration. Of these, SPECTRAL TILT measures have been the most popular, and have been shown to characterize phonation contrasts in many languages. We use spectral tilt measures based on differences in amplitudes between individual harmonics in the spectrum.

Figure 2. Sample harmonic spectrum with labels for the four lowest-frequency harmonics (H1, H2, H3, H4), the harmonic nearest 2000 Hz (“2k”), and harmonics nearest the formant frequencies F1, F2, and F3 (A1, A2, A3). Since the spectrum shows only up to 3000 Hz, the harmonic nearest 5000 Hz is not shown. The frequency of the first harmonic (H1) is the fundamental frequency ($f_0$).

Figure 2 shows a harmonic spectrum in which several individual harmonics are labeled (H1 etc.). (The lowest-frequency harmonic is the fundamental component, and its frequency is the fundamental frequency, $f_0$.) The amplitude of each harmonic is its height in the vertical dimension. In the spectrum of voicing at the larynx, without the influence of the vocal tract, harmonic amplitude is expected to decline with frequency. Spectral tilt, or roll-off, refers to the steepness of that decline. A spectrum with strong higher-frequency harmonics has little tilt, and
the high-frequency energy makes the voice sound brighter and tenser/more pressed. In contrast, a spectrum with weak higher-frequency harmonics has a greater tilt, and the lack of high-frequency energy makes the voice sound breathier and weaker. Spectral tilt matters not only over the spectrum as a whole, but within particular sub-ranges of frequencies, especially at the low-frequency end of the spectrum, where the amplitude of the first harmonic is very important. Sundberg and Gauffin (1979) found that the absolute amplitude of the first harmonic reflects the amount of air flowing through the glottis (“The greater the maximum airflow amplitude in the flow glottogram, the stronger the voice source fundamental.”). A less constricted glottis during voicing means greater airflow and a stronger H1. At the same time, less glottal constriction results in a more sinusoidal vibratory pulse, which has relatively more energy in the fundamental than in higher harmonics. Thus with a more open glottis, H1 both absolutely and relatively has a higher amplitude. The higher-frequency part of the spectrum, in contrast, is thought to depend more on the abruptness of closing of the vocal folds as they vibrate.

While the amplitudes of the individual harmonics can be calculated, these measures depend in large part on the overall amplitude of a sound, and so most research instead uses amplitude differences between pairs of harmonics. For example, we work with the measures H1−H2, H2−H4, H4−H2k, H2k−H5k, H1−A1, H1−A2, and H1−A3. The first four of these measures together approximate, piecewise linearly, the gross overall shape of the harmonic spectrum between the fo and 5000 Hz. Kreiman and colleagues (e.g. Kreiman & Gerratt 2011, Garellek et al. 2016, Kreiman et al. 2021) have shown that these four measures characterize well the spectral differences across individual voices, that they can vary independently of one another, and that all of them matter perceptually to listeners. In terms of articulation, harmonic source spectral slope measures correlate with glottal open/contact quotient and medial vocal fold thickness (for a recent overview of the relationship between articulation and commonly-used acoustic measures of phonation, see Garellek 2019.)

A source spectrum may also include inharmonic components. Both turbulent flow in the glottis (essentially, aspiration), and aperiodic vibration of the folds (e.g. jitter or shimmer) add noise in the sense of reducing the signal-to-noise ratio (Kreiman and Gerratt, 2005). Such noise components of the source can vary in overall spectral shape, but at this time it is not known how sensitive listeners are to such spectral differences, and therefore it is not known how the noise
spectrum should be modeled and measured (though see Kreiman et al. 2021 for one model of the noise spectrum). Some common measures of noise in the source spectrum compare the amplitudes of the harmonic and noise (inharmonic) components: these are called HARMONICS-TO-NOISE RATIO measures. The value of a harmonics-to-noise ratio is affected by the levels of both the harmonics and the noise. A high value indicates that the vibration is strongly periodic, with strong harmonics, and that there is relatively little noise of any kind. A low value indicates either that the harmonics are weak — the vibration is not strongly periodic, or with little overall energy — or that there is significant noise of some kind. Thus, modal phonation will show the highest values, while both prototypically creaky and breathy phonations will show lower values.

In our work we use the VoiceSauce analysis program (Shue 2010, Shue et al. 2011; available for free download at [http://www.phonetics.ucla.edu/voicesauce/](http://www.phonetics.ucla.edu/voicesauce/)), to make these and other measures automatically over many audio files. Only with such batch processing is extensive cross-language comparison possible. Possible advantages of using VoiceSauce rather than Praat for this purpose are discussed in Shue et al. (2011).²

The proliferation of acoustic measures, especially those that can be made easily and automatically by VoiceSauce, Praat, or other programs, can be confusing. Brunelle & Kirby (2016:200) consider the “variable utility of different voice quality measures” to be a pressing phonetic issue for the study of tone in languages. Some researchers have directly compared measures in how well they distinguish different contrastive phonation categories. For example, Blankenship (2002) compared many measures (mostly made by hand) and found three that were the most useful (H1−H2, H1−A2, and Cepstral Peak Prominence, a harmonics-to-noise ratio measure). Using automatic measures provided by VoiceSauce, Esposito (2012) found that CPP, H1*, H2*, and H1* − H2* all show statistically significant differences between phonation types in Hmong. Nonetheless, in our study we include almost all the measures provided by VoiceSauce; we can then determine which acoustic measures are most useful in defining the cross-language space using data reduction and correlation techniques, as explained in §2.4.

1.3. RESEARCH QUESTIONS. What acoustic properties structure the space of phonation types across languages? One possible answer is that the space is comprised of a single dimension, perhaps corresponding to a Ladefoged-style continuum of glottal airflow, as described above
(Ladefoged 1971, Ladefoged & Maddieson 1996). Yet as also noted above, Ladefoged’s continuum did not include strident/harsh phonation types, which presumably require an additional dimension.

In contrast, a more recent suggestion in the literature about individual speaker voice differences is that at least five dimensions are needed. The voice source model of Garellek et al. (2016) mentioned above proposes four frequency bands for harmonic energy differences, plus glottal noise (measured in four frequency bands in Kreiman et al. 2021). While these dimensions clearly underlie differences between voices of individual speakers, it is not necessarily the case that the same dimensions characterize different types of phonation found across languages. Indeed, it has been suggested that the acoustic space for linguistic phonation might be simpler than the cross-speaker acoustic space (Garellek et al., 2013); after all, the entire population of speakers, with their individual physiological differences, must be able to produce linguistic contrasts. In this study, we focus on the cross-language space for different vocalic phonation types across languages, with the expectation that fewer than five dimensions are likely to suffice.

If we have an overall acoustic space for phonation type across languages, we can then ask how different languages use this space. One possibility is that in every language, distinct phonation types (whether contrastive or allophonic) are maximally dispersed, always using the full range of available phonatory qualities, regardless of the number of phonation categories. Alternatively, languages with more categories could use an expanded phonetic space, beyond the range used by languages with fewer categories. These alternative scenarios are already familiar in phonetics from considerations of cross-language vowel and consonant spaces (Becker-Kristal 2010, Schwartz et al. 1997a, Lindblom & Maddieson 1988).

Finally, we can ask how similar are the “same” categories or contrasts across languages. The IPA treats only three categories of phonation: modal, breathy, and creaky voice, with the latter two marked by diacritics. There are also strong traditions describing lax/slack voice, tense/stiff/pressed voice, and strident voice, even though the IPA does not include diacritics for these. (Unofficial extensions of the IPA, as well as ad hoc transcription systems for individual languages, include several proposals for how to represent these phonation types.) Whenever a small number of phonetic categories is applied to many different languages, these realizations are bound to differ. We know that, for example, breathy vs. modal contrasts are made in different
languages in slightly different ways acoustically: for example, Esposito (2010a) compared
breathy and modal vowels in 10 languages on seven different acoustic measures and found that
no two languages made the contrast in exactly the same way. Similarly, Keating et al. (2015)
provide acoustic criteria that distinguish several kinds of creaky voice, and it is possible that
different languages could use different subsets of these kinds (though see Kim et al. 2020 for
similarities in creak realization in Mandarin, Cantonese, and Hmong). In general, it would not be
surprising if different languages have somewhat different realizations of their phonation types;
phonetic categories of all kinds are never exactly the same across languages. For example,
probably no two languages with vowels transcribed with the same phonetic symbols have exactly
the same phonetic vowel qualities; and no two languages have exactly the same distributions of
Voice Onset Time (VOT) values. It is very unlikely that phonation categories would be any
different from these more familiar kinds of phonetic categories. Nonetheless, just as with these
more familiar categories, we expect instances across languages that have been given the same
phonetic label to cluster together in the larger phonetic space. The question then is how tightly
they will cluster — that is, how similar the categories are when looking across languages.

2. METHODS

2.1. LANGUAGES AND PHONATION CATEGORIES. In this study we compare the contrastive or
allophonic phonations of 11 languages, from five language families across different continents;
many of them are understudied. In part the choice of languages was opportunistic, based on
availability of suitable audio recordings. Other languages were then selected so that the sample
would better represent the typology of phonation contrasts described above, as follows.

As described above, the basic glottal-airflow phonation continuum of voicing is taken to
comprise three major categories — modal, breathy, and creaky voice — plus two intermediate
categories, tense voice (between creaky and modal) and lax voice (between breathy and modal).
We wanted our sample to include languages with modal/breathy/creaky contrasts, as well as
languages with tense/lax contrasts. In addition, so that noisy phonation types are represented, one
of the languages in our sample is !Xóô, with harsh (‘strident’) and pharyngealized voice qualities
includes examples of seven phonation categories. We also included two languages with no phonation contrast.

Modal voice is by definition the most common phonation type in the world, and it is also the most common in this language sample: all but the three Yi languages (i.e., Southern Yi, Bo, Hani) have this category. English here is coded only for this category; no effort was made to identify individual English tokens with creaky voice, which in this language is a feature with primarily prosodic, paralinguistic and/or sociolinguistic functions (Podesva & Callier 2015, Crowhurst 2018), rather than being phonemic or allophonic. Breathy voice occurs contrastively in Gujarati, Hmong, Miao, Mazatec, !Xóõ, and Zapotec. Creaky voice is contrastive in Mazatec, !Xóõ, and Zapotec, and occurs allophonically with some lower-pitched tones in Hmong, Mandarin, and Miao. The three Yi languages contrast tense vs. lax voice, and Miao has an allophonically tense-voiced high tone (Kuang 2013). In Mazatec and !Xóõ, a creaky-voice high tone vowel sounds tense rather than creaky, but is coded here as creaky. The categories distinguished within each language, contrastively and allophonically, are given in Table 1. Sample audio files for most of these languages will be included in the Supplemental Materials.

It would be possible to limit this sample to phonation categories that are clearly contrastive in each language, but instead we also include some cases (labeled “allophonic”) in which the phonation property seems to be a secondary correlate of a contrast. We include these cases because the phonation appears fairly consistently and is clearly audible (and even salient to native speakers), and because the weighting of a primary vs. a secondary correlate of a contrast is gradient — in one language a secondary correlate might be quite minor, but in another language it might be almost as informative as the primary correlate. See Kuang and Cui (2018) and studies by Abramson and colleagues cited elsewhere here for cases of shifting cue weights in register contrasts, and see Brunelle and Kirby (2016) for general discussion of multiple correlates of register and tone contrasts. That is, it is not always clear when a phonetic feature or correlate is being used contrastively in a language, so our choice of cases was liberal.

Nonetheless, we label each case as either contrastive or allophonic, so here we discuss our use of these labels. In many languages tone and phonation cross-classify in the lexicon (that is, combine more or less freely on a single vowel); a well-known and striking example is Mpi as exemplified by Harris and Ladefoged
Here, our examples are the Yi languages (Bo, Hani, Southern Yi), Mazatec, and !Xóõ. In such cases it is clear that all the phonation differences are contrastive, since they cannot be predicted from the tone or pitch.

Table 1. Languages, phonation categories, and speakers in this study.

<table>
<thead>
<tr>
<th>Language (Abbrev.) (Family)</th>
<th>Phonation categories</th>
<th># Speakers (F, M)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bo (Bo) Sino-Tibetan</td>
<td>Contrastive: Lax, Tense</td>
<td>10 (4, 6)</td>
</tr>
<tr>
<td>English (En) Indo-European</td>
<td>Modal</td>
<td>18 (11, 7)</td>
</tr>
<tr>
<td>Gujarati (Gj) Indo-European</td>
<td>Contrastive: Breathy, Modal</td>
<td>10 (7, 3)</td>
</tr>
<tr>
<td>(White) Hmong (Hm) Hmong-Mien</td>
<td>Contrastive: Breathy, Modal, Allophonic: Creaky</td>
<td>23 (8, 15)</td>
</tr>
<tr>
<td>(Luchun) Hani (Lu) Sino-Tibetan</td>
<td>Contrastive: Lax, Tense</td>
<td>9 (4, 5)</td>
</tr>
<tr>
<td>(Black) Miao (Mi) Hmong-Mien</td>
<td>Contrastive: Breathy, Modal, Allophonic: Creaky, Tense</td>
<td>8 (0, 8)</td>
</tr>
<tr>
<td>Mandarin (Mn) Sino-Tibetan</td>
<td>Modal</td>
<td>31 (16, 15)</td>
</tr>
<tr>
<td>(Jalapa) Mazatec (Mz) Otomanguean</td>
<td>Contrastive: Breathy, Creaky, Modal</td>
<td>12 (6, 6)</td>
</tr>
<tr>
<td>!Xóõ (Xo) Tuu</td>
<td>Contrastive: Breathy, Creaky, Harsh, Modal, Pharyngealized</td>
<td>12 (0, 12)</td>
</tr>
<tr>
<td>(Southern) Yi (Yi) Sino-Tibetan</td>
<td>Contrastive: Lax, Tense</td>
<td>12 (6, 6)</td>
</tr>
<tr>
<td>(Valley) Zapotec (Zp) Otomanguean</td>
<td>Contrastive: Breathy, Creaky, Modal</td>
<td>6 (2, 4)</td>
</tr>
</tbody>
</table>
Furthermore, if a language has two phonations which occur on the same pitch level or pitch contour, or one phonation which occurs on two different pitch levels or contours, then the phonations likewise are contrastive, since again they cannot be predicted from the pitch. Our examples of this are the breathy phonations of Miao and Hmong (in Miao, there are modal and breathy mid tones; in Hmong there are modal and breathy high-falling tones), and the creaky phonation of Zapotec (there are creaky low-falling and high-falling tones).

The four phonation instances classified as allophonic in Table 1 involve co-variation with pitch as part of a tonal category. In general, phonation is expected to vary somewhat across a voice’s pitch range: to reach the voice’s lowest or highest pitches, the vocal folds must vibrate somewhat differently. For example, this is known to be the case in English (e.g. Kuang 2013). As a result, tones with extra-low or extra-high pitch often have a concomitant phonatory quality, predictable from the pitch. A well-known example is that in Mandarin the low-pitched portions of tones 2 (213) and 4 (51) are usually creaky, though occasionally breathy (Kuang 2017: 1700; Moisik et al. 2014). Since this creaky voice is fairly consistent, we code these two Mandarin tones as a separate Creaky category; and since it is predictable from the low pitch, we consider it to be allophonic. Similarly, in White Hmong (Esposito 2012, Garellek 2012) and Black Miao (Kuang 2013) the lowest-pitched tone is usually creaky (though in Xinzhai Hmu (Miao) the low tone is described as breathy, Liu et al. 2018), and these low tones also are listed as instances of allophonic creak. It is noteworthy in this regard that Garellek et al. (2013) showed that only breathy voice is criterial for native White Hmong listeners’ tone recognition, while creaky voice is not attended to, in accord with our classification of its breathy phonation as contrastive, but its creaky phonation as allophonic.

Finally, in Black Miao, a language with five level tones (Kwan 1966, Kuang 2013), the pitch range is expanded on the high end as well as the low end; the highest tone, with an extra-high pitch, has a tense phonatory quality. We consider this to be an instance of allophonic tense phonation. Thus Miao, with its large number of level tones and its expanded pitch range, is treated as using allophonic creaky voice for its lowest tone and allophonic tense voice for its highest tone, in addition to contrastive breathy and modal voice on its mid tones.
2.2. Wordlists and Audio Recordings. For each language with phonation contrasts, a set of words was compiled to provide several minimal- or near-minimal pairs for the contrasts. The numbers of such pairs differed across languages. The words were mostly constrained to be monosyllables (or disyllables, in the case of !Xóō) and to exclude high vowels, nasalized vowels, nasal consonants, and voiceless fricatives, as these segment types present challenges to automatic acoustic analysis of voice quality.

The languages without phonation contrasts, English and Mandarin, did not follow these constraints, because available corpora used voiceless fricative and nasal onset consonants. These corpora consisted of one or two CV sequences rather than many different words: for English, the word “sure”, and for Mandarin, minimal sets formed with [ma] and [ʂə] with the four tones. This second Mandarin vowel has a first formant frequency well above the fundamental frequency, and with widely spaced F1/F2/F3.

For the tone languages, all of their level tones, plus some or all of their contour tones, were included in the wordlists, though their numbers were not balanced. To provide some pitch variation, the English recordings selected included both statement and question intonations. Because the Gujarati words were recorded in unique sentences (see below), they also exhibit some pitch variation.

For the languages other than Mazatec and !Xóō, speakers were recorded using either a computer soundcard, or PCQuirer (Scicon Research & Development Inc.) with its external D/A box, all at a sampling rate of 22kHz, with a head-mounted Shure SM10A unidirectional microphone (50–15000 Hz frequency response) close to the corner of the mouth. Recordings of four indigenous languages of southwestern China (i.e. Bo, Hani, Miao, Yi) were made in the field.5 Recordings of Hmong and Zapotec were made in language community centers in the US, and recordings of English and Gujarati were made in our laboratory soundbooth. Our Mandarin corpus combines recordings made in Beijing and in our laboratory soundbooth. For Mazatec and !Xóō, existing recordings were accessed from the online public UCLA Phonetic Archive (http://archive.phonetics.ucla.edu/); the recordings available there are by Peter Ladefoged with Paul Kirk and Tony Traill, respectively. In this kind of sample, differences in recording conditions are unavoidably confounded with language differences. We mitigate this confound by standardizing the resulting measurements by speaker, rather than analyzing the raw data.
Table 1 also shows the number of speakers analyzed for each language. While nine languages have both men and women speakers, two languages have only men, and overall it was not possible to evenly balance across gender. In general, we expect women’s voices to be breathier than men’s voices, but specifically with respect to the phonetic realization of phonation categories, large gender effects are not expected for these languages. This is because previous studies have found that in Mazatec, Gujarati and Hmong there were no gender effects on phonation (Garellek and Keating 2011, Khan 2012, and Esposito 2012 respectively), while in our Yi languages (Bo, Hani, Southern Yi) gender differences are seen only in secondary aspects of the physiological electroglottographic signal (Kuang and Keating 2014), while Kuang (2011) found no acoustic effects in Southern Yi. Notably, Esposito (2011) found gender differences in a different Zapotec language, but we do not have a large enough sample here to know if those differences hold across Zapotec varieties. Thus the expectation is that gender differences will be ones of scaling (with women breathier overall), and this kind of difference is controlled for by standardizing measurements within speakers, and comparing only these standardized measures across speakers. For this reason, we do not think that the different numbers of men vs. women for the different languages in our sample will account for apparent differences between languages.

For most of the languages, test words were spoken in isolation; in Hmong, test words were spoken in a short carrier sentence. Gujarati speakers are prone to produce spelling pronunciations of words with breathy vowels in formal settings and when reading, so we followed the elicitation procedure described in Khan (2012), in which written prompts were shown only briefly, after which speakers were asked to quickly create their own sentences beginning with the test words and produce them as many times as possible within a 10-second window.

In some of the languages, simultaneous electroglottographic recordings were made for the larger project; however, these will not be discussed in the present paper.

2.3. ACOUSTIC MEASURES. Audio recordings were converted to WAV format for analysis; Version 23 (July 5, 2015) of VoiceSauce, precompiled for standalone use on PCs, was used. A preliminary step with VoiceSauce is to segment and label the vowels of interest from the test utterances, using Praat to make TextGrid files. A Praat script can be used to help with this task,
but to some extent this is done manually. Then VoiceSauce is run to estimate a set of acoustic parameters for each of these labeled segments.

In the version of VoiceSauce used here, the original Kawahara STRAIGHT algorithm (Kawahara et al. 1998) estimates the fundamental frequency at 1-msec intervals. Harmonic spectra are computed pitch-synchronously over windows of three pitch pulses. Given the $f_0$ estimate, VoiceSauce uses an optimization function to locate the harmonics of the spectrum, and finds their amplitudes. This method greatly reduces variability compared to methods that use a fixed-length window. (Note that $f_0$ is calculated by VoiceSauce here only as the basis for finding the harmonics; $f_0$ itself is not included in any of our analyses of phonation quality shown here, as preliminary analysis revealed that, not surprisingly, it modeled tone categories instead of phonation categories.) For voiceless intervals, which can sometimes occur in non-modal vowels with particularly strong constriction or aspiration, VoiceSauce outputs no $f_0$ or harmonic amplitudes.

VoiceSauce then uses the Snack Sound Toolkit (Sjölander 2004) to find the frequencies and bandwidths of the first four formants, also at 1-msec intervals. The harmonics nearest to these formant frequencies are located, and their amplitudes are taken as the amplitudes of the formants. Finally, the formant frequencies, along with stored estimates of their bandwidths, are used in an algorithm that corrects harmonic amplitudes for the filtering effects of the vocal tract, using Iseli et al.’s (2007) extension of Hanson’s (1995) method.6 (In uncorrected measurements, the formant frequencies boost the amplitudes of any nearby harmonics, and since different vowel qualities have different formant frequencies, this very large effect of the filter can dwarf any harmonic amplitude differences in the source.) Corrected harmonic amplitudes are indicated by an asterisk, e.g. $H_1^*$.7 There may also be interactions between vowel quality and phonation type on harmonic amplitudes, even for corrected measures; Esposito et al. (2019) found that vowels with higher formant values were associated with lower $H_1^*-H_2^*$. However, overall, these corrections allow us to combine data from different vowel qualities.

In the subsequent analysis, we also include measures of the individual harmonic amplitudes, because these can also be associated with differences across phonation types; for example, Esposito (2012) found that $H_1^*$ was better at distinguishing modal and creaky phonation types than $H_1^*-H_2^*$.
Other VoiceSauce parameters index periodicity and/or noise using cepstral analysis.\textsuperscript{8} Harmonics-to-noise ratios (HNRs) over four frequency bands (0–500 Hz, 0–1500 Hz, 0–2500 Hz, 0–3500 Hz) have high values for very periodic signals with strong harmonics, and/or when spectral noise is low; in VoiceSauce they are computed using de Krom (1993)’s algorithm. Cepstral Peak Prominence (CPP; Hillenbrand, Cleveland, & Erickson 1994) is an instance of HNRs over the entire frequency range, in which the cepstral peak is normalized relative to a regression line in the cepstrum between 1 msec and the maximum quefrency; VoiceSauce uses Hillenbrand et al.’s algorithm. Sun (2002)’s Subharmonic-to-Harmonic ratio (SHR) describes the relative strength of any subharmonics (interharmonics) in the spectrum. Subharmonics in the spectral domain correspond to alternating periods in the time domain, and thus SHR indexes period doubling, which often occurs in creaky voice and with broader laryngeal constriction.

Finally, pitch-synchronous Energy, and Strength of Excitation (Murty & Yegnanarayana 2008), are calculated. Strength of Excitation (a measure available in VoiceSauce since 2015 but not frequently used) reflects the relative amplitude of impulse-like excitation in each pitch pulse. It is generally correlated with overall Energy, but depends more on the glottal excitation and less on vocal tract effects or noise. With greater glottal constriction, voicing weakens. Thus, Garellek (2020) found that the constricted phonation types in !Xóõ (creaky, pharyngealized, and harsh) are characterized by lower SoE relative to phonation types (like breathy and modal) with little constriction.

VoiceSauce produces an output text file that gives the mean value of each parameter for each labeled segment; in this study we also output means over each third of each labeled segment interval. We note that languages with phonation contrasts can have distinctive timing patterns, with non-modal phonations occurring on only portions of the vowels (Silverman 1997, Blankenship 2002, Garellek et al. 2021). In this paper, we focus on the phonation types occurring in the middle third of the vowel, since preliminary analysis showed that this interval was most informative across the languages, presumably because it avoids coarticulatory effects from neighboring consonants, and because non-modal phonation is generally strong over this interval (even if it is stronger earlier or later).

All values from VoiceSauce were then z-score standardized within speaker. Standardizing first within speaker avoids modeling speaker-specific (idiosyncratic) or gender-specific patterns
instead of language x phonation categories. The mean standardized values (averaging across individual tokens) of the middle-third interval of each acoustic measurement were then calculated for each language x phonation category. As will be seen below, one of our analyses uses this reduced language x phonation dataset, while the other analysis uses the individual data tokens.

2.4. **Statistical analysis: Multi-Dimensional Scaling and Classification and Regression Trees.** The multiple acoustic measures made in this study can be thought of as defining a multi-dimensional acoustic space within which different phonations can be located. However, many of these measures are inter-correlated, so the dimensions of the space are not necessarily independent, and can readily be reduced in number without much loss of information. Our goal is a map of the low-dimensional space that best fits the acoustic data, a map in which phonation categories can be located such that distances between the categories reflect their (dis)similarities. Multi-Dimensional Scaling (MDS; Kruskal & Wish 1978) is one method for reducing many individual measures to a smaller number of independent dimensions.9 MDS is a class of analysis that aims at reducing distances between categories in a multi-dimensional space, with the specific goal of mapping distances or dissimilarities between categories along a smaller number of independent dimensions. MDS is particularly suited to dealing with data with strong multi-collinearity, as is found among our voice measures. And, because MDS does not assume linearity, it is preferred over Principal Components Analysis, another method of data reduction. However, MDS analysis of hundreds or thousands of data points, as we have, does not produce meaningful results. Therefore this analysis uses the language x phonation means because we are interested in the distances between the phonation categories across the languages, not in the individual speakers or tokens.

Clopper (2012) gives examples of MDS in linguistic phonetic research; generally, these involve perception data, either confusion matrices or similarity judgments. Our use is somewhat different, since our data are from production; we take the sizes of acoustic differences as distance (dissimilarity) measures.

MDS uses measured distances between items to define a map in which those distances are preserved in a lower-dimensional space. The Manhattan (or city-block) distances on the set of
acoustic measurements were used as the basis for estimates of the physical distances between all pairs of language x phonation categories, and these distances were input into Kruskal’s Non-metric Multidimensional Scaling algorithm (performed in R using the metaMDS function in the vegan package). Non-metric MDS is used for non-Euclidean distances, as we have here.

The algorithm yielded solutions with different numbers of dimensions, where more dimensions typically do better at preserving the original distances, but too many dimensions offer diminishing returns in data-fitting, and can be hard to interpret and visualize. To evaluate how well a particular low-dimensional space reproduces the original distance matrix, stress values were calculated as a measure of goodness-of-fit. The smaller the stress value, the better is the fit of the reproduced space. A scree plot (Figure 3) was plotted to visualize the stress values against different numbers of dimensions. The “elbow” point of the curve was identified as the sufficient number of dimensions. As suggested by Figure 3, a two-dimensional space is able to sufficiently reproduce the original distance matrix. However, the third dimension adds some additional information that helps to highlight some of the subtler phonation contrasts. The dimensions of an MDS solution can then be explored by correlating them with the original acoustic measures.

Figure 3. Scree plot for the MDS solution.
In order to have a better understanding of how phonation categories are classified by voice parameters, taking advantage of measurements of all individual tokens, a Classification And Regression Tree (CART) (Breiman et al. 1984) was fitted to the full dataset of standardized measures from the vowel middle thirds, using the `rpart` package in R. This analysis thus complements our MDS analysis. The depth of the tree was determined by a complexity parameter, which was optimized through a process of cross-validation and grid search. The purpose of this procedure is to find the tree that has the best predicative accuracy. The complexity for the final model was kept at the default, restrictive, value of 0.01.

3. Results

3.1. 2-D Space. The MDS 2-dimensional solution can be viewed as a spatial map, a kind of voice space for phonation distinctions. This solution is shown in Figure 4, in which each datapoint is an average phonation category in one language. Each phonation category is plotted in a different color, the same across languages: breathier phonations are shown in shades of orange (breathy = dark orange, lax = light orange), creakier phonations are shown in shades of blue (creaky = dark blue, tense = light blue), modal phonation is shown in green, and the other two phonations are purple (pharyngealized) and pink (harsh).

First, we consider the overall organization of this space. !Xóõ, the language with the most phonation contrasts in this dataset, defines the extreme edges of the space: Each dimension makes a two- or three-way distinction in !Xóõ, with its Modal (green) and Creaky (dark blue) maximally distant on Dimension 1, and its Breathy (dark orange) maximally distant from Creaky (dark blue) on Dimension 2. The categories of the other languages lie within the diamond-shaped space defined by !Xóõ. English Modal (green), the only category coded for that language, lies right in the center of the space.

Second, we consider the two dimensions individually. Dimension 1 (x-axis) seems to give a rough Modal (on the left) to non-Modal (on the right) continuum: Modal (green) points lie mostly to the left, while the points with other colors lie mostly to the right. Modal (green) and Harsh (pink) points are well-distinguished on this dimension. Tense (light blue) and Lax (light orange) points lie together in the middle, where Modal (green) and Breathy (dark orange) points also overlap. Dimension 2 (y-axis) seems to roughly represent a harmonic-spectral-tilt
continuum, from more high-frequency energy (orange-family points, at the bottom) to less high-frequency energy (blue-family points, at the top). Breathy (dark orange) and Creaky (dark blue) points are well-distinguished on this dimension. Tense (light blue), Lax (light orange), and Modal points (green) lie in the middle, overlapping with some of the Creaky (dark blue) points. Harsh phonation (pink) lies with the Breathy (dark orange) points, but also near the Creaky (dark blue) points. Pharyngealized phonation (purple) is like Creaky (dark blue) here, on both dimensions.

Figure 4. 2-dimensional MDS solution for the 29 language x phonation categories from the 11 languages, each category a set of acoustic measures from the middle third of all vowels. Color coding: green = Modal (M), dark orange = Breathy (B), dark blue = Creaky (C), light blue = Tense (T), light orange = Lax (L), purple = Pharyngealized (P), pink = Harsh (H). Language
Third, we consider whether the various instances of the “same” category tend to cluster together. We expect this to be more or less the case because the same category names are used across languages, but it is certainly possible that any single datapoint might look more like instances of some other category, because the voice space could be divided up and labeled differently in descriptions of different languages.

In Figure 4, three major clusters can be seen by focusing on the different-color points. In the upper-to-middle right corner are the Creaky (dark blue) and Pharyngealized (purple) points. At the bottom are the Breathy (dark orange) and Harsh (pink) points. The Tense (light blue), Lax (light orange), and Modal (green) lie mostly in the center of the space, but extend out to the far left. The Lax (light orange) points are closest to Breathy (dark orange), while the Tense points (light blue) are closest to Creaky (dark blue). The Tense (light blue) and Lax (light orange) points mostly form tighter clusters within this middle region, though Miao Tense lies a bit apart.

Modal (green), in contrast, is more spread out: there is a general Modal area, but !Xóõ Modal lies at some distance to the left. Also, some instances of Modal are more like other categories. Notably, Mazatec Modal patterns with the Breathy tokens, but this is presumably because the Mazatec Breathy−Modal contrast is most prominent during initial third of the vowel, not during the middle third as analyzed here (Garellek & Keating 2011).

Otherwise, the obvious outliers are Gujarati Breathy (dark orange) and the Zapotec Creaky (dark blue). There is greater similarity between the Gujarati Breathy and the Lax (light orange) phonation in other languages, and greater similarity between the Zapotec Creaky and Tense (light blue) phonations.

It is entirely possible that these characterizations are phonetically accurate: the Gujarati Modal−Breathy contrast could be characterized as a Modal−Lax contrast, while the Zapotec Modal−Creaky contrast could be characterized as a Modal−Tense contrast. Indeed, the relative distances between these categories in this 2-D space are fairly short.

Fourth, we can compare the 11 languages in terms of how well their phonation categories are separated in the space. Visually, we can think of this as connecting all the datapoints for a
language, making a triangle/rectangle etc, much in the same way as vowel spaces are described. How big is this language-specific space? We might expect a dispersion factor (e.g. Lindblom & Maddieson 1988, Schwartz et al. 1997) to be at work here: the more categories a language contrasts, the more the language should use this 2-D space, until that space is saturated. Figure 5 facilitates such comparisons by using a different color for each language; other than the color-coding, the information is exactly the same as in Figure 4. As already noted, !Xóõ (brown points) is almost optimal, distinguishing four of its categories in a big square that uses both dimensions of the space. (The remaining category distinction will be discussed below).

Two languages have three or four categories (allophonic or contrastive), which visually form “triangles” within the 2-D space. Hmong (light green), with three categories, is well-dispersed. Three of Miao’s four categories (purple), including its allophonic Creaky and Tense along with its contrastive Breathy, cover an area similar in size to Hmong’s, but then Miao’s Modal category sits in the middle of this triangle.
Figure 5. 2-dimensional MDS solution for the 29 language x phonation categories from the 11 languages, each category a set of acoustic measures from the middle third of all vowels. Same as Figure 4 except that here, colors code languages rather than phonation categories. Color coding: brown=!Xóõ, red=Bo, black=English, yellow=Gujarati, light green=Hmong, blue=Luchun Hani, fuschia=Mandarin, grey=Mazatec, purple=Miao, dark green=Yi, turquoise=Zapotec

Two languages have three categories falling on a line. Zapotec’s points (turquoise) lie along a line defined by the two dimensions together, but the categories are not as well-separated as in other languages, due to Zapotec Creaky lying in the Modal area. The three categories of Mazatec (grey) lie on a line defined mostly along only Dimension 2.
The languages with two categories do not use the 2-D space as fully. The Yi languages (red, blue, and dark green) and Gujarati (yellow) — which in fact also patterns like the Yi languages — use primarily Dimension 2 for their contrasts, with relatively small distances between their categories. On the other hand, Mandarin (fuchsia), with its two allophonic categories, keeps them well separated along only Dimension 1. This is the only language that uses Dimension 1 as its primary dimension. Also, as previously noted, English (black) lies in the middle of the space, making no contrasts.

Figure 6. Pair-wise 2-dimensional spaces involving the third dimension of the MDS solution. The left panel shows Dimension 1 vs. Dimension 3, while the right panel shows Dimension 2 vs. Dimension 3. The color coding of the phonation categories and the language x phonation codes are the same as in Figure 4.

What about !Xóõ’s fifth category? It was noted above that Dimension 3 of the MDS solution does add some additional information beyond what the first two dimensions provide. Figure 6 shows the two pairwise 2-D spaces involving the third dimensions of the solution. Here, Dimension 3 is on the y-axis. It can be seen that on this dimension, !Xóõ Creaky (dark blue) and Pharyngealized (purple) are well distinguished, though in a way that enhances the difference
between !Xóô Creaky and the cluster of other Creaky datapoints, which !Xóô Pharyngealized is close to. Dimension 3 also mildly enhances the difference between Tense (light blue) and Lax (light orange) datapoints on Dimension 2. That is, it takes three dimensions to fully differentiate the five phonation categories of !Xóô.

Last, we can test which acoustic measures are making the dimensions of the space, by checking the weights of the measures on each dimension. Table 2 shows how 21 acoustic measures from VoiceSauce relate to each of the three dimensions. Dimension 1, which looks like a non-modal to modal continuum, is most strongly based on measures of periodicity and energy in the excitation or the speech signal – SHR, the Strength of Excitation, and Energy – but also on H4*–H2k*, a mid-frequency spectral tilt measure. CPP and the other HNRs – measures of periodicity and noise – also contribute to Dimension 1, but not as strongly. Dimension 2, which looks like a glottal airflow continuum, is based on measures complementary to those for Dimension 1: unsurprisingly, H1*–H2* and H1*–A1* – measures of low-frequency spectral balance – but also, perhaps surprisingly, SHR, a measure of periodicity. Dimension 3, which does relatively little work in distinguishing the categories, is also based on periodicity – SHR – though presumably in a way different from on Dimension 2.

This division of labor between the parameters across the dimensions is also made clear by running MDS analyses on subsets of parameters. When only the harmonic-amplitude measures are used, the first dimension of the MDS solution gives a spectral-tilt continuum, like Dimension 2 above. When the other measures are used instead, the first dimension of that MDS solution gives a modal vs. non-modal distinction, like Dimension 1 above.

It is interesting that SHR, a measure proposed by Sun (2002) to reflect period-doubling in creaky voice, is strongly weighted on all three dimensions. The pattern on Dimension 3, where !Xóô Creaky is distant from the other Creaky categories while !Xóô Pharyngealized is near them, suggests that most Creaky categories and !Xóô Pharyngealized show period-doubling, while !Xóô Creaky, like most Modal categories, does not. Nonetheless, it turns out that, overall, removing SHR from the set of acoustic measures submitted to MDS does not change the overall shape of the solution and its dimensions.
Table 2. Weight of each acoustic measure on each dimension of the 3-D MDS solution (D1, D2, D3). Parameters that have higher weights than other parameters on each dimension are in boldface (weights > 2.0 for D1 and D2, and the highest-weighted parameter for D3).

<table>
<thead>
<tr>
<th>Measure</th>
<th>D1</th>
<th>D2</th>
<th>D3</th>
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<td></td>
<td></td>
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<tr>
<td>H2*--H4*</td>
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<tr>
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<tr>
<td>H4*--H2kHz*</td>
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</tr>
<tr>
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<tr>
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<tr>
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</tr>
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<tr>
<td>Root mean squared (RMS) Energy</td>
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<td>0.87</td>
</tr>
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</table>

3.2. How subsets of languages use the space. From the solution for the 11 languages presented above, it is not clear how the individual languages in the sample, and the phonation categories each contributes to the overall set, influence this overall solution. We can get a better
We have already seen that !Xóõ is the main determinant of the structure of the 11-language space. When that language is left out (bottom left graph in Figure 6), the resulting space is like a
zoomed-in view of the other languages. Hmong, Mazatec, and Miao Breathy/Creaky/Modal form the edges of this new space. The Miao categories are now better separated, with its Modal joining the English Modal in the center. The Tense–Lax distinctions are clearer.

English remains in the center not only in the 11-language space, but also in all of the 10-language subsets. However, in many spaces derived from smaller subsets (not seen in Figure 7), English appears in odd places – in a corner with other, contrastive Modal categories, or with other categories entirely.

When MDS is run on only the languages with Tense–Lax contrasts, the resulting space (shown in the lower-right corner in Figure 7) distinguishes their categories much more clearly. For this subset, the new Dimension 1 is related primarily to $H1^*-H2^*$. Conversely, when these languages are omitted, the solution for the remaining languages is essentially unaffected. That is because the Tense–Lax contrasts are not important in structuring the voice space for all 11 languages.

3.3. CLASSIFICATION TREE. The analyses above, by looking at weights of measures on dimensions, and spaces defined by subsets of measures, give a fair idea of how different measures contribute to making different category distinctions within a voice space. As described above, an analysis that focuses on the relative importance of the various acoustic measures, and which uses all of our individual data points, comes from a Classification And Regression Tree (CART) analysis (Breiman et al. 1984). The classification tree obtained from the rpart package in R is shown in Figure 8. (In Figure 8 and throughout, “B” stands for breathy, “M” for modal, and “C” for creaky categories.) The classification tree is then paired with boxplots of distributions of values on each measure that appears in the tree, here separated by the 29 language x phonation categories. These are shown in Figures 9a-c.
Figure 8. Classification tree of phonation categories derived from the full set of acoustic measures (middle thirds). The colors here are arbitrary, unrelated to those in previous figures. Key: “B” = breathy, “M” = modal, “C” = creaky, HNR = Harmonics-to-noise ratios, SHR = Subharmonic-to-Harmonic ratio, and H1H2 = the difference in amplitude between the first and second harmonics.

Figure 8 shows that the first division is made by the HNR between 0–500 Hz, which is by definition doing the most work, splitting off most of the Modal categories from the non-Modal ones (reminiscent of Dimension 1 in the MDS analysis). Figure 9a shows the distributions of values on this measure across language x phonation categories. It can be seen that Modal categories generally have higher values on this measure, though there is a spread, and that it seems to best distinguish Modal from Creaky categories.

In Figure 8, the non-Modal categories are then split by SHR into Breathy vs. non-Breathy. As was seen in Table 2, SHR weights strongly on all three dimensions of the MDS voice space. Figure 9b shows the distributions for this measure. Breathy categories mostly have the lowest values, while most of the Creaky categories, and Pharyngealized, have higher values. It is interesting here that SHR mostly contributes to the distinction between Breathy and other phonation types, even though (as reviewed earlier) it was developed as a way to track period-doubled creaky voice. It seems likely that Modal voice often is somewhat creaky (which contributes to the presence of stronger sub-harmonic energy), in contrast to Breathy voice, which has very little sub-harmonic (or even harmonic) energy.
Finally, in Figure 8 the non-Breathy remainder is divided into Creaky vs. (residual) Modal by $H1^*-H2^*$. Figure 9c shows generally declining values from left to right, and thus reflects a spectral tilt continuum. (Obvious exceptions are that Gujarati Breathy again is more like Lax or Modal than like the other Breathy categories, as is Harsh voice compared to Creaky phonation.) With respect to the classification tree, we can see in Figure 9c that this measure is generally lower for Creaky. This classification tree does not show enough splits to divide Tense or Lax phonations from Modal, or Harsh or Pharyngealized from other phonations.

It can be seen in Figs. 9a-c that some phonation x language categories are much more variable (across speakers and tokens) than others on a given measure. Most strikingly, Gujarati in all three panels, Hmong Modal in Fig. 9-b, and Hmong Breathy and Modal in Fig. 9-c show many outlier datapoints. Our best guess as to why this is, is that these are the two languages where the test words were recorded in sentences, which could well contribute more variability. In contrast, there is no apparent pattern in variability based on contrastive vs. allophonic phonation, or number of contrastive categories. Notably, English Modal shows fairly tight distributions, even though with no contrast it is free to cover a range of phonations, as indeed Mandarin Modal does.
Figure 9a. HNR 0–500 Hz z-scores, middle third of vowels. Colors and language x phonation codings as in Figure 4 above.
Figure 9b. Subharmonic-to-Harmonic Ratio z-scores, middle third of vowels. Colors and language x phonation codings as in Figure 4 above.
4. DISCUSSION AND CONCLUSIONS.

4.1. DIMENSIONS OF THE SPACE. The analyses presented above indicate that the space for the phonations of our sample of 11 languages is largely 2-dimensional, as presented in Figure 4. Within this space, Creaky and Pharyngealized categories cluster together in one region, Breathy and Harsh categories cluster together in an adjacent region, and Modal (and Modal-like) categories are spread out in the rest of the space. We note that a separate MDS analysis of just the male speakers from each language shows a very similar result, with slightly better clustering of the phonation categories, but the same overall organization.

The structure of the space bears on the question of the status of “Modal” as a phonation category: Is there indeed a distinction between modal and non-modal phonations? Our results indicate that there is, but it is a continuous rather than a binary distinction. The first (and therefore by definition most important) dimension of the space distinguishes Modal phonations
from Creaky, Pharyngealized, and Harsh, with Tense and Lax at the non-modal edge of the Modal categories, and Breathy similar to Tense and Lax.

On the other hand, the second dimension of the space bears on the question of the status of a phonation type continuum, as popularized by Ladefoged. This dimension is indeed structured like a traditional Breathy-to-Creaky glottal airflow continuum. Here Modal phonations are in the middle; Tense is between Modal and Creaky, while Lax is between Modal and Breathy. This dimension appears to be basic to phonation contrasts. First, the languages contrasting just two categories (the Yi languages and Gujarati) mainly use this dimension for their distinctions. Second, even Mazatec, with three contrasting categories, mostly uses only this dimension. Third, all languages with three or more categories use this dimension. That is, there is no language that doesn’t use this dimension for at least some phonemic contrast.

Nonetheless, !Xóõ’s contrasting categories are distinguished only when all three dimensions are considered. This case provides an example of dispersion within the voice space as a function of the number of contrasts. !Xóõ has more contrasts than the other languages in our sample, and it not only needs all three dimensions to make these contrasts, but it also makes the most use of the 2-dimensional space. Lindblom and Maddieson (1988) proposed that increasing numbers of contrasting segments first lead to expanding and filling a basic phonetic space, up to a point where the space itself must become more complex by incorporating additional dimensions of contrast. Other languages are not as dispersed in the 2-D space, and the languages with just two categories mostly use just a single dimension (Dimension 2), and are not very separated even along that one. The English non-contrastive Modal category is in the middle of the space, not pushed to the modal edge. We have already suggested that the Breathy-to-Creaky phonation continuum is the most basic space for phonation contrasts; the modal/non-modal distinction is the first expansion of the space beyond that, and a third dimension is also available for even further expansion.

Thus, our results indicate that the phonetic space for phonation is more complex (of higher dimensionality) than the Ladefoged-style continuum, but at the same time it is less complex than the voice source model of Garellek et al. (2016). As previously suggested, it does seem that the space for linguistic contrasts is simpler than the space for individual speaker differences.
4.2. Further sub-types. We noted in the Introduction that sub-types of both Creaky and Breathy phonations are attested, but the linguistic significance of these sub-types is unknown. Do our results contribute to this issue? In our data, the answer to this question seems partly related to the answer to another question we posed: whether the contrastive vs. allophonic status of the use of non-modal phonation matters.

Recall that we classified four categories as allophonic: Creaky in Mandarin (with no phonation contrasts) and in Hmong and Miao (with contrastive Breathy and Modal), plus Tense in Miao (also with contrastive Breathy and Modal). The allophonic Tense category of Miao patterns just like the contrastive Tense categories of the three Yi languages. However, the results for the Creaky voice categories are different. The three allophonic Creaky cases (Mandarin, Hmong, Miao) do pattern differently from the contrastive cases (Mazatec and !Xóõ). These allophonic distinctions are made along Dimension 1, with these Creaky categories in the Modal region on Dimension 2. The Miao and Hmong cases are more extreme – they lie at one end of Dimension 1, while Mandarin Creaky is closer to Modal on that dimension. (This difference cannot be due to Mandarin Creaky being more modal, since if anything Mandarin Tone 3 and Tone 4, especially for Beijing speakers, have more creak mid-vowel than do the Miao and Hmong low tones.) In contrast, Mazatec and !Xóõ Creaky are Modal-like on Dimension 1 but lie at an extreme on Dimension 2, the Breathy-to-Creaky dimension, even though both of these languages locate the creak more at vowel edge. As discussed in the Results section, all five Creaky categories lie in a separate part of the overall space (in the upper right corner), but we can speculate that contrastive vs allophonic status (Mandarin, Hmong, Miao vs. Mazatec and !Xóõ) might determine the kind of Creaky voice used in each language. Other than these differences, we see no evidence for different kinds of Creaky voice across languages (in the sense of Keating et al. 2015).

A possible reason for the importance of Dimension 1 in some languages’ Creaky category is found in Huang’s (2020) study of the role of individual phonation cues in Mandarin tone perception. Huang found that Mandarin perception depends on the low fo and/or irregular phonation properties of creak, but not on its harmonic spectrum properties. In our analysis, irregular phonation (periodicity) is most strongly related to Dimension 1, while the harmonic spectrum is related more to Dimension 2; that is, the information about Mandarin creak is in
Dimension 1, not Dimension 2. That is, production and perception are aligned in Mandarin; speakers use the acoustic cues that listeners attend to.

With respect to distinguishing breathy from whispery voice within our broad Breathy category, things are less clear. Hmong, Mazatec, Miao, and Zapotec are fairly close together in the space, while !Xôõ is more extreme and Gujarati less extreme on Dimension 2. If we accept the idea that Harsh voice in !Xôõ is whispery because the vocal folds are more spread while the epilarynx is constricted (as shown in Traill 1986 and argued recently by Garellek 2020), then we look for Breathy cases lying near our one Harsh case in order to identify a whispery part of the space. Miao and Zapotec are less modal on Dimension 1, nearer to the Harsh case, than are Hmong and Mazatec; but the difference is so small that it is at best suggestive.

4.3. ACOUSTIC PARAMETERS. Our results have also shown which acoustic parameters seem most important in structuring this space. Certain measures are most strongly correlated with the dimensions of the MDS solution: for Dimension 1 (roughly modal to non-modal), SHR, Strength of Excitation, and Energy; for Dimension 2 (roughly glottal constriction), H1*–H2*, H1*–A1*, and again SHR; for Dimension 3 (a different modal to non-modal spread), again SHR. On the other hand, the splits on the classification tree (where all individual datapoints are used) are made firstly by HNR 0–500 Hz (for Modal vs. others), SHR (for Breathy vs. others), and H1*–H2* (for Creaky). SHR and H1*–H2* are thus the parameters that are important in both analyses, and HNR 0–500 Hz is most important in the classification tree analysis, where it seems to take the place of MDS Dimension 1’s SHR. These three parameters reflect different aspects of the voice source spectrum, and therefore different aspects of phonation. Higher H1*–H2* is correlated with decreased vocal fold contact during voicing (DiCanio 2009) as well as greater medial fold thickness (Zhang 2016). SHR measures the strength of subharmonics in the spectrum, and thus higher values index irregular vocal fold vibration (Sun 2002, Keating et al. 2015). A lower harmonics-to-noise ratio is correlated with increased aspiration and, especially in the lower frequencies as here, voicing irregularity. Of the other measures here, Strength of excitation is a measure of the strength of voicing; increased constriction at the glottis or in the larynx is associated with weaker voicing (Chong et al. 2020, Garellek 2020), whereas higher H1*–A1* may additionally reflect greater posterior glottal opening (Hanson et al. 2001).
A major, and surprising, result here is the dominance of the SHR across analyses and dimensions. Its role here clearly goes well beyond indexing period-doubled creaky voice. Indeed, in these analyses of our language sample, it is not especially connected to creaky voice at all.

We have noted that the Gujarati Breathy category lies with the Yi languages’ Lax categories, rather than the Breathy categories of other languages. This may be related to the fact that Gujarati contrasts Breathy vowels vs. Modal vowels preceded by the aspiration of a Breathy consonant; Esposito & Khan (2012) demonstrated that vowels in this latter category have much stronger acoustic and electroglottographic cues to breathiness, suggesting that contrastive Breathy vowels in Gujarati are characterized by a subtler degree of breathiness in comparison. Notably, this result is seemingly at odds with Tian & Kuang’s (2019) study of breathy voice in Shanghainese and three of the languages from our sample (Hmong, Gujarati, Southern Yi). They compared the importance of different acoustic parameters in characterizing breathy vowels in each language, and found that Gujarati was more like Hmong than like Southern Yi: while all three languages favored harmonic amplitude correlates, only Southern Yi had virtually no noise correlate. (Shanghainese was more different yet, with an almost entirely noise-based category.) A possible explanation for these opposite results lies with the more extensive set of acoustic parameters used in the present study. Tian & Kuang (2019) did not include SHR in their model, yet in our results, that one parameter is important in structuring the overall phonetic space. We know very little about this parameter, and relatively little voice research that uses VoiceSauce has included it. Clearly, however, it merits detailed future study aimed at understanding exactly what it tells us about phonation.

These results, then, offer a perspective on the acoustic parameters that could be selected for including in linguistic studies of phonation. Researchers may find the array of parameters in VoiceSauce overwhelming. In our own past practice, we have tended to select one or two harmonic measures, plus CPP as a noise measure. Now, with this new, more inclusive, analysis, we can recommend $H_{1^*} - H_{2^*}$, SHR, HNR 0–500 Hz, and optionally $H_{1^*} - A_{1^*}$, Strength of Excitation, and Energy, as a small set of the most informative parameters. These measures also capture differences in the way phonation types are articulated: $H_{1^*} - H_{2^*}$ and $H_{1^*} - A_{1^*}$ are associated with vocal fold contact, medial fold thickness, and posterior glottal opening; SHR is associated with period doubled phonation, and HNR 0–500 Hz with low-frequency noise (such
as from irregular voicing or aspiration); SoE is associated with voicing intensity, and Energy with overall sound intensity. Other measures apparently are either not very different between phonation categories, or are so highly correlated with one of the measures just listed that they provide little additional information.

We conclude this section with a reminder about the limitations of this study. While a corpus of 11 languages goes well beyond the typical single-language study of phonation, it is nonetheless a very small sample for a typological or general-phonetic study. We look forward to future research that will expand cross-language comparisons to dozens, or hundreds, of languages. (See footnote 10 for tools that could facilitate this.) As mentioned earlier, differences among our language samples with respect to recording conditions, speaker gender balance, and the words recorded (e.g. different vowel qualities, different tones, in a sentence or not) could confound what appear to be language differences. Our results also depend on the validity and accuracy of our measurements, and as with any automated measurements, the data here may be noisy, especially the corrected harmonic amplitude measures. All of these factors have the potential to introduce errors and random variation in the estimation of acoustic measures and consequently in the statistical analysis. However, such errors are likely to be unsystematic, such that they could affect data from any phonation type, speaker, or language. Therefore, we believe such errors are unlikely to change the phonation spaces that were obtained in this study.

4.4. IMPLICATIONS FOR PHONOLOGICAL REPRESENTATIONS. We have shown, not surprisingly, that when many acoustic measures are made of a phonatory distinction, (1) multiple measures will contribute to making the distinction, but (2) some of these will be more important (play a larger role) than others. We have also shown how acoustic similarity among phonation types can be visualized in a low-dimensional phonetic space, as is routinely done (using raw acoustic measures already known to be important) for e.g. vowel quality and Voice Onset Time. The kind of data reduction and weighting that we have illustrated here for phonations can help phoneticians make sense of complex phonetic categories and contrasts, identifying probable primary vs secondary cues. Can it also be of any relevance for phonologists? If we take 2-D vowel formant plots for comparison, we can see that the two dimensions of such plots correspond to sets of phonological features crucial to vowel behavior: one dimension shows
vowel height, which corresponds to two or more phonological height/tension features, while the other dimension shows vowel front-backness, which in most languages corresponds to a single phonological backness feature. The more categories lie along a phonetic dimension, the more features are needed to distinguish those categories; for example, if a language also has vowel rounding contrasts, then there will be more phonetic categories along the (primarily) front-back dimension, requiring an additional feature.

Our 2-D plot of the vowel phonation space can, speculatively, be viewed in a similar way. We have interpreted Dimension 1 in this space as distinguishing Modal from non-modal categories. This suggests a possible feature [+Modal], which would group together seemingly-disparate categories like Creaky and Breathy as laryngeally marked. We do not know if in any languages with phonation contrasts these categories pattern together as a class in opposition to Modal. However, Gallagher (2012 and elsewhere) provides a similar case from Quechua, where ejective and aspirated stops (with constricted and spread glottis, respectively) pattern together in opposition to plain stops. For phonation, we can offer the hypothesis that such a pattern is also possible.

We have interpreted Dimension 2, on the other hand, as a phonetic continuum that ranges over several categories from Creaky to Breathy. This then suggests features like those for vowel height, perhaps [+Breathy] and [+Creaky], with Modal being [-Breathy, -Creaky]. It is an open question whether Tense and Lax phonations would be [+Creaky] and [+Breathy] respectively, with a reduced phonetic realization, or whether an additional feature is needed. That will depend in part on whether any languages are found to contrast Breathy vs. Lax, and Creaky vs. Tense.

4.5. CONCLUSIONS. In sum, our study has demonstrated an overall acoustic space for voice quality. We looked at a diverse range of languages with contrasting or allophonic phonation categories, and found that most such categories can be accommodated within a 2-dimensional space, though the most complex system uses 3 dimensions. Just as with the more familiar phonetic category distinctions of vowel quality and VOT, phonation categories that are given the same labels in different languages are found in similar parts of the overall space, but nonetheless differ across languages.
APPENDIX

Further information about the 61 languages with phonation contrasts on vowels mapped in Figure 1.

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<tr>
<td>Zapotec, Yalálag</td>
<td>Otomanguean</td>
<td>M, C</td>
<td>Y</td>
<td>Avelino (2010)</td>
</tr>
<tr>
<td>Zhuang, Du'an</td>
<td>Tai-Kadai</td>
<td>M, C</td>
<td>Y</td>
<td>Perkins et al. (2016)</td>
</tr>
<tr>
<td>ǂHoan, Eastern</td>
<td>Kx'a</td>
<td>M, C, P</td>
<td>Y</td>
<td>Honken (2013)</td>
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<tr>
<td>!Xóõ (Taa)</td>
<td>Tuu</td>
<td>P, H, B, M, C</td>
<td>Y</td>
<td>Trail (1985); Naumann (2016); Garellek (2020)</td>
</tr>
</tbody>
</table>
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NOTES

1 In the recent modelling work by Zhang (2016), variation in glottal opening is less important than medial vocal fold thickness in causing changes to voice acoustics. The distance between the arytenoids is important only with respect to noise generated in the glottis.

2 These advantages of VoiceSauce hold for ports from Matlab to other languages, such as Octave or Python (https://github.com/voicesauce), but not for analyses done entirely within Praat, such as by PraatSauce (https://github.com/kirbyj/praatsauce)

3 But see the extended clinical VQ system of Ball, Esling, & Dickson 2018.

4 Esling (2005) describes the tense/lax contrast in another Yi language as a difference in vowel quality; see Kuang and Cui (2018) for discussion of ongoing sound changes in the Yi languages, and for evidence that the speakers in our sample produce a phonation contrast.

5 We thank Professor Jiangping Kong of Peking University for his assistance with these fieldwork trips and recordings, and for permission to use the Hani recordings made during one trip.

6 Local correction of harmonic amplitudes has come to replace inverse filtering as the way to minimize filter effects; see Gobl and Ni Chasaide (2012) for review. The traditional alternatives to some such correction are: (1) match the vowel qualities (and thus formant frequencies) of the stimulus sets to be compared, or (2) analyze only vowels with a high first-formant frequency, using only acoustic measures below this high F1 (e.g. H1-H2 measured for low vowels).

Even with these corrections, we do try to avoid high vowels in our sample; in automatic formant estimation, the very low F1 of most high vowels may not be correctly separated from the fundamental for women speakers, and corrections made with faulty formant estimates will give faulty harmonic amplitude estimates.

7 Simpson (2012) shows that H1-H2 can pattern quite differently for men’s vs women’s voices independently of e.g. breathiness level, because the sexes tend to differ in degree of nasality, and
nasality affects harmonic amplitudes. H1*-H2* is equally problematic for this reason, since the effects of nasality are not estimated in the all-pole LPC filter model and thus not corrected for. Simpson thus cautions against using H1-H2 measures to compare the sexes.

8Traditional measures of voicing irregularity measured in the time domain, such as jitter and shimmer, are not included in VoiceSauce because these are not perceived independently of spectral noise; see Kreiman & Gerratt 2005.

9For a quite different approach to comparing phonetic categories across languages, in which failures of an automatic classifier to distinguish categories index their similarity, see Thomson, Nearey and Derwing (2010).

10We have developed a Shiny app for this project, available from https://pennplab.shinyapps.io/MDS_v1/, which facilitates MDS and boxplots of a dataset. Our own dataset, available from https://www.dropbox.com/s/m6bz9mhrnh5jviq/lg11_all_color.csv?dl=0, can be re-analyzed with different subsets of languages or measures; anyone can add their own new data to the dataset, and see how the results change.