Does the Auditory System Efficiently Code All Languages Or Just American English?

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INTRODUCTION

Lewicki (2002) used Independent Component Analysis (ICA) to examine statistical properties of human speech. Statistically optimal filters for encoding speech were well-aligned with frequency tuning in the mammalian auditory nerve (comparing measures of Q_{10}), leading Lewicki to suggest speech makes efficient use of coding properties of the auditory system. However, these analyses only examined American English, which is neither normative nor representative of the world's languages. Here, ICA was used to compare optimal encoding of speech from 14 different languages found across the world with physiological response properties.

METHODS

Stimuli

Recordings of 14 languages (Dutch, Flemish, Greek, Javanese, Jul'hoan, Norwegian, Swedish, Tagalog, Tahitian, Urhobo, Vietnamese, Wari', Xhosa, Yeyi) were collected, mostly from the UCLA Phonetics Lab Archive (<u>http://archive.phonetics.ucla.edu/</u>). All recordings were at least one minute long and contained clear speech tokens from a native speaker without any background noise. Recordings were high-pass filtered at 125 Hz and divided into 8-ms samples (after Lewicki, 2002).

ICA

In ICA, the observed data **x** are assumed to be the result of linear combinations of s:

$$\mathbf{x} = \mathbf{A}\mathbf{s}$$

where **A** is a mixing matrix whose columns constitute basis functions, and s is a source vector with components s_i that are statistically independent from each other. A and s are unknown, so ICA estimates them as follows:

$$\mathbf{v} = \mathbf{W}\mathbf{x}$$

[2]

W is an unmixing matrix of the same dimensionality as A ($W = A^{-1}$), making the output y the recovered source vector which approximates \mathbf{s} up to scaling and permutation. The rows of W are statistically optimal filters for recovering source signals \mathbf{s} from the observed mixtures \mathbf{x} .

Maximum likelihood ICA was used (Pearlmutter & Parra, 1996) with the natural gradient extension to facilitate convergence. W was iteratively updated by stochastic gradient descent: [3]

$$\mathbf{W} = [\mathbf{I} - \operatorname{sign}(\mathbf{y})\mathbf{y}^{\mathrm{T}}]\mathbf{W}$$

where **I** is the identity function and $sign(\cdot)$ is the sign function. W is initialized to the identity matrix, and ΔW is the change in the unmixing matrix that is added to W at each iteration. ICA was conducted for 20,000 iterations, with a different batch of 500 samples randomly selected for analysis at each iteration.

Regression Analysis

When ICA is complete, each row in W is a statistically optimal filter for encoding input stimuli. Sharpness of each filter (Q_{10}) was calculated when possible (when filter response decreased by 10 dB above and below the center frequency). Linear regressions were calculated for Q_{10} as a function of center frequency on a log-log scale, following Lewicki (2002).

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