

# A method for automatically detecting problematic tongue traces

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While ultrasound provides a remarkable tool for tracking the tongue’s movements during speech, it has yet to emerge as the powerful research tool it could be. A major roadblock is that the means of appropriately labeling images is a laborious, time-intensive undertaking. In work reported at ICPR in 2010, Fasel and Berry (2010) introduced a “translational” deep belief network (tDBN) approach to automated labeling of ultrasound images. The current work extends that methodology with a modification of the training procedure to reduce reported errors (Sung and Archangeli, 2013) along the anterior and root edges of the tongue by altering the network’s loss function and incorporating  $\ell^1$  and  $\ell^2$  regularization (Ng, 2004) to avoid overfitting. This training-internal approach to error reduction is compared to an independent post-processing procedure which uses the expected average positional change between adjacent points in three tongue regions (Davidson, 2006) to detect and constrain erroneous coordinates. Positional variance was calculated using the 800 most diverse and 50 least diverse tongue configurations by image pixel intensity across multiple subjects from a recitation of the phonetically balanced Harvard sentences (Rothausser et al., 1969).

**Index Terms:** articulatory phonetics, ultrasound imaging, tongue imaging, speech processing, deep belief networks, regularization, computer vision

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