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Sensorimotor Representation of Speech Perception — Cross-Decoding of Place of Articulation Features during Selective Attention to Syllables in 7T fMRI

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35 Abstract

36	Sensorimotor integration, the translation between acoustic signals and motoric programs, may
37	constitute a crucial mechanism for speech. During speech perception, the acoustic-motoric
38	translations include the recruitment of cortical areas for the representation of speech
39	articulatory features, such as place of articulation. Selective attention can shape the processing
40	and performance of speech perception tasks. Whether and where sensorimotor integration
41	takes place during attentive speech perception remains to be explored. Here, we investigate
42	articulatory feature representations of spoken consonant-vowel syllables during two distinct
43	tasks. Fourteen healthy humans attended to either the vowel or the consonant within a syllable
44	in separate delayed-match-to-sample tasks. Single-trial fMRI BOLD responses from
45	perception periods were analyzed using multivariate pattern classification and a searchlight
46	approach to reveal neural activation patterns sensitive to the processing of place of
47	articulation (i.e., bilabial/labiodental vs. alveolar). To isolate place of articulation
48	representation from acoustic covariation, we applied a cross-decoding (generalization)
49	procedure across distinct features of manner of articulation (i.e., stop, fricative, and nasal).
50	We found evidence for the representation of place of articulation across tasks and in both
51	tasks separately: for attention to vowels, generalization maps included bilateral clusters of
52	superior and posterior temporal, insular, and frontal regions; for attention to consonants,
53	generalization maps encompassed clusters in temporoparietal, insular, and frontal regions
54	within the right hemisphere only. Our results specify the cortical representation of place of
55	articulation features generalized across manner of articulation during attentive syllable
56	perception, thus supporting sensorimotor integration during attentive speech perception and
57	demonstrating the value of generalization.

58 Significance Statement

59	Speech is supported by sensorimotor integration, a bidirectional translation of its auditory and
60	motoric signals. Whether our brain represents speech as articulatory features during selective
61	attention has not yet been well specified. We focused on the representation of articulatory
62	information of speech during attentive speech perception. For the first time, we applied
63	generalization in classification analysis to counteract the differences in acoustic properties
64	that accompany articulatory information. Participants attended to either the vowels or
65	consonants of syllables, while undergoing fMRI. Our results show that articulatory
66	information is represented in widespread cortical areas during selective attention to the
67	different syllable components, supporting sensorimotor integration during attentive speech
68	perception.

69 Introduction

70 Speech is supported by sensorimotor integration, a (bidirectional) translation of its auditory 71 and motoric signals. These translations, which occur during speech perception (Hickok and 72 Poeppel, 2007; Evans et al., 2016; Schomers and Pulvermüller, 2016), can result in the 73 cortical representation of articulatory features of speech, such as place of articulation, manner 74 of articulation, and voicing. Particularly, the cortical representation of place of articulation 75 features has been reported in dorsal speech regions, including motor and premotor areas 76 (Pulvermuller et al., 2006), and somatosensory and supramarginal regions under passive 77 listening (Correia et al., 2015). However, differential task requirements can modulate cortical 78 representations of articulatory features. For example, variation of somatotopic activations in 79 motor areas were found during passive sound perception involving different articulators 80 (Pulvermuller et al., 2006a). Other researchers have reported differential patterns in superior 81 temporal but not in the motor cortex during an incidental task with phonemes (Arsenault and 82 Buchsbaum, 2015). Several reasons have been discussed to account for this variability; among 83 them, differences in task demands across studies (e.g., type of task, number of items, but also 84 passive vs. active tasks, i.e., selective attention) as well as third factors such a covariation of 85 manner of articulation during place processing.

Although the underlying neural mechanisms of speech perception and attention remain elusive, they are explained in terms of network dynamics (Friederici and Singer, 2015) and, in particular, of theta-gamma amplitude or phase coupling (Giraud and Poeppel, 2012; Hyafil et al., 2015) of neural activity. For example, attention can implement phase resetting and entrainment of neuronal oscillations to a relevant stimulus stream (Lakatos et al., 2008) and can rapidly change the spectrotemporal receptive field to enhance task-relevant stimulus properties (Fritz et al., 2003). Importantly, selective attention has been shown to generate Ultra high-field 7 Tesla fMRI allows investigating the living human brain with unprecedented
high spatial resolution (Yacoub et al., 2001), signal-to-noise ratio (Vaughan et al., 2001), and
specificity (Polimeni et al., 2010). Beyond measurement improvements, multivariate pattern
analysis (MVPA) further increases the sensitivity of experimental contrasts by exploiting
concurrent spatial patterns of fMRI responses (Norman et al., 2006). Crucially, MVPA has
allowed unraveling information representation of abstract speech features (Correia et al.,
2015; Evans and Davis, 2015).

102 Here, we aim to minimize the effect of covariation in manner of articulation during place of 103 articulation processing by generalization across manner. We study the cortical representation 104 of place of articulation features of syllables using 7T fMRI and MVPA-based cross-decoding 105 (i.e., generalization). Specifically, we exploited the acoustic variation imposed by two 106 manners of articulation (e.g., stop and fricative) to identify patterns discriminative of place of 107 articulation (i.e., bilabial/labiodental versus alveolar) and tested these patterns in a third 108 (unseen) manner of articulation (e.g., nasal). This procedure capitalized on the acoustic 109 variation imposed by different manner of articulation features to extract specific patterns for 110 place of articulation features (Ladefoged and Johnson, 2010).

This generalization was applied to two different tasks with identical auditory stimuli (i.e., attention to consonants and attention to vowels) to investigate the neural representation of place of articulation features during selective attention to speech and the possible effects of attention on the neural representation. Previous studies have shown representation of place of articulation features in sensorimotor regions (Correia et al., 2015) and a modulatory role of attention in sound and phonetic representations of speech (Bonte et al., 2014; Downer et al.,

117 2015). Based on these studies, we used decoding of place of articulation features as a metric 118 of sensorimotor processes during speech perception. We then investigated whether decoding 119 was possible on either task and, if so, whether sensorimotor integration is modulated by the 120 selective attention to different syllable components (i.e., vowel or consonants). The idea 121 behind choosing vowels and consonants tasks was that place of articulation is more 122 representative for consonant identification, and less for vowel identification in Consonant-123 Vowel structures. This orthogonal stimuli-task arrangement allowed assessing the effect of

124 attention on the place of articulation representation without explicit task focus to place.

125 Materials and Methods

126 Participants

- Fourteen native Dutch speakers (mean \pm SD age, 27.2 ± 5.1 years; 9 females; 2 left-handed)
- 128 underwent blood oxygenation level-dependent (BOLD) signal fMRI scanning while
- 129 performing two delayed-match-to-sample tasks. Participants had no history of neurological or
- 130 systemic diseases and reported normal hearing abilities. All participants received monetary
- 131 compensation for their participation and signed written informed consent. The Ethical
- 132 Committee of the authors' Home Institution approved the study.

133 Stimuli

134 Stimuli were 18 consonant-vowel (CV) syllables recorded by 3 female native Dutch speakers,

135 generating 54 unique tokens. The syllables were the result of the CV combinations of six

- 136 consonants (p, t, f, s, m, n) and three vowels (a, i, u). This subset of consonants was selected
- 137 because it allowed us to build a balanced stimuli matrix that covers two features of place of
- 138 articulation (i.e., bilabial/labiodental and alveolar) and three features of manner of articulation
- 139 (i.e., stop, fricative, and nasal). This arrangement was needed to perform MVPA-based cross-
- 140 decoding analysis and to keep the necessary number of trials for single-trial classification (see

141 Multivariate analysis below). Syllable recordings were selected from a subset of recordings 142 previously used in our laboratory (Correia et al., 2015). Briefly, the consonants composing the 143 syllables for every articulatory feature were bilabial/labiodental, ('p', 'f', 'm') and alveolar ('t', 's', 'n') for place of articulation; and stop ('p', 't'), fricative ('f', 's'), and nasal ('m', 'n') 144 145 for manner of articulation (Figure 1A). The three different vowels and three different speakers 146 introduced acoustic variability useful for classification. All stimuli were recorded in a 147 soundproof chamber at a sampling rate of 44.1 kHz and digital-to-analog converted with 16-148 bit resolution. Stimuli were presented in the MRI scanner via MR-compatible earphones with 149 a linear frequency transfer of up to 8 kHz (Model S14, Sensimetrics Corporation, USA). 150 Before starting the experiment, the volume was adapted to each subject's audible and 151 comfortable perceptual level.

152 Experimental design

153 Participants performed two delayed-match-to-sample tasks in four runs of 54 trials each (i.e., 154 each syllable token was presented once per run) during fMRI acquisition. In cognitive terms, 155 this task allowed paying attention to a certain aspect of the acoustic stimuli – from now on 156 referred to as "attend to vowels" and "attend to consonants." The task also involved 157 remembering and matching the attended vowel or consonant later on with a visually presented 158 token (i.e., a written vowel or consonant) of the same category (with 50% match/mismatch 159 response proportions). Each run was divided into two blocks of 27 trials, one for each task, 160 and the tasks where counterbalanced across runs and participants. During the attend to vowels 161 task, participants heard a syllable and received, after 6-8 seconds, a written vowel as "match 162 cue." Participants were instructed to match the vowel of an auditorily presented CV syllable 163 with a written vowel cue as fast and accurately as possible by pressing a button with the right 164 index ('match') or middle ('mismatch') finger. During the attend to consonants task, 165 participants heard the same syllables (of the attend to vowels task) and received a written

166

167 Importantly, the features of interest (i.e., place of articulation) were never part of an explicit 168 task to avoid confounds in attentional demands. The delay-match-to-sample task is widely 169 used to precisely control the subject's attention. Thus, data were analyzed only for the first 170 part of each trial before the cue onset to avoid that relevant signal was contaminated with 171 matching and response related processing period signals.

consonant as cue that should be matched with the auditory syllable (see Figure 1B).

172 The task was clearly explained to each subject outside the scanner, and they received at the 173 beginning of each block an introduction display indicating the syllable component they should 174 attend to (i.e., vowels or consonants) at the beginning of each task. A trial consisted of (1) a 175 speech stimulus sound presentation (i.e., a CV syllable token of 340 milliseconds duration) 176 followed, after an inter-stimulus interval (ISI) of 6-8 seconds, by (2) a written cue (i.e., a 177 consonant or vowel letter matching or not the sound token depending on the specific block 178 and trial of 2 seconds duration, written in Times New Roman, font size 30, black color); and 179 then followed by (3) the subject's 'match'/mismatch' immediate response. The sound 180 syllables and the visual cue (i.e., a written vowel or consonant) matched in 50% of the trials, 181 and 'match'/'mismatch' trials were balanced across attention conditions and randomized per 182 subject. Mismatch cues were always of the same category (i.e., vowels in the attend to vowels 183 task and consonants in the attend to consonants task). In total, each run lasted 15 minutes and 184 the behavioral responses were collected through an MR-compatible button box (Current 185 Designs, 8-button response device, HHSC-2x4-C; Philadelphia, USA).

We used fast sparse image acquisition to have a 500 ms silent gap to present each syllable
sound (Di Salle et al., 2003). The fMRI acquisition was set up as a slow event-related design.
The inter-trial interval (ITI) between consecutive auditory stimuli was relatively long (i.e., 14
seconds on average; range: 12 to 16 seconds) to allow independent BOLD signal estimation
per trial (see Figure 1C). The interval used to estimate the fMRI response to the spoken

syllable perception per trial was restricted to the first 6 seconds following the sound
presentation to avoid contamination from the processing of the visual cue, matching, or button
press. Moreover, as the acoustic, phonetic, and phonological features of the presented
syllables were identical across trials and attention conditions, the effects on the cortical
representations pertain to differences in attention. All stimuli, event identities, and timings
were presented and logged using Presentation from Neurobehavioral Systems
(www.neurobs.com, RRID: SCR_002521).

198 Functional MRI acquisition

199 Functional and anatomical volumes were acquired on a whole-body Siemens Magnetom 7 200 Tesla scanner (Siemens, Erlangen, Germany) and a 32-channel head-coil (Nova Medical Inc.; 201 Wilmington, USA) at the author Home Institution Imaging Center. For all functional runs, we 202 acquired whole brain high-resolution accelerated multiband gradient echo (T2*-weighted) 203 echo-planar imaging (EPI; Moeller et al., 2010; Setsompop et al., 2012) data (echo time, TE = 204 21 ms; repetition time, TR = 2,000 ms; time of acquisition, TA = 1,500; delay in TR (silent 205 gap = 500 ms; multi-band factor = 3; generalized auto-calibrating partially parallel acquisitions (GRAPPA) g-factor = 2; flip angle = 72° ; field of view, FOV = 198 mm; voxel 206 size = $1.5 \times 1.5 \times 1.5 \text{ mm}^3$; number of slices = 72, without gap between slices) for each 207 208 participant. To correct for EPI geometric distortions, 10 volumes with opposite phase 209 encoding directions (i.e., posterior to anterior and anterior to posterior) were additionally 210 acquired using the same acquisition parameters as in the functional runs. After the first 2 211 functional runs, we acquired a tridimensional T1-weighted magnetization prepared rapid 212 acquisition gradient echo (3D-MP2RAGE; Marques et al., 2010) volume (240 sagittal slices; voxel size = $0.65 \times 0.65 \times 0.65 \text{ mm}^3$; first inversion time TI1 = 900 ms; second inversion time 213 214 TI2 = 2,750 ms; TE = 2.51 ms; TR = 5,000 ms; first nominal flip angle = 5°; second nominal 215 flip angle = 3°) per participant.

216 Functional MRI preprocessing

217	Anatomical and functional data were analyzed using BrainVoyager QX (version 2.8.4; Brain
218	Innovation; Maastricht, Netherlands, RRID: SCR_013057), and custom code written in
219	MATLAB (R2014a version 8.3.0.532; The Mathworks Inc.; Natick, MA, USA, RRID:
220	SCR_001622). Anatomical images were interpolated to a nominal voxel size of 1.5 x 1.5 x 1.5
221	mm ³ matching the functional images' resolution. The functional images were corrected for
222	motion artifacts using the 3D rigid body motion correction algorithm implemented in
223	BrainVoyager QX and all functional runs were aligned to the first volume of the second
224	functional run. We corrected the EPI distortions using the topup tool implemented in FSL
225	(RRID: SCR_002823, (Smith et al., 2004). The reversed phase encoding images, acquired
226	after the anatomical images, were used to estimate the susceptibility-induced off-resonance
227	field and, then, to correct the distortions in the remaining functional runs. After this
228	correction, functional data were high-pass filtered using a general linear model (GLM)
229	Fourier basis set of eleven cycles sine/cosine, including a linear trend removal. Functional
230	volumes per run were co-registered and aligned to the anatomical scan using rigid body
231	transformations (i.e., six parameters: three translations and three rotations). Finally, functional
232	images were normalized by transformation into Talairach space (Jean and Tournoux, 1988).
233	Anatomical segmentation and cortex-based alignment. Intensity inhomogeneities in T1-
234	weighted images were first corrected using Statistical Parametric Mapping 12 (SPM12,
235	RRID: SCR_007037 Ashburner and Friston, 2005) software. The resulting images were used
236	to perform volumetric segmentation with the FreeSurfer analysis software (version 5.3.0,
237	http://surfer.nmr.mgh.harvard.edu/, RRID: SCR_001847) (Dale et al., 1999). Briefly, this
238	processing includes motion correction, removal of non-brain tissue using a hybrid
239	watershed/surface deformation procedure, automated Talairach transformation, intensity

240 normalization, tessellation of the gray/white matter boundary, automated topology correction,

241 and surface deformation following intensity gradients to optimally place the gray/white and 242 gray/cerebrospinal fluid borders at the location where the greatest shift in intensity defines the 243 transition to the other tissue class. Quality control was performed by visually inspecting each 244 subject's brain after the process was finished. Remaining errors were manually corrected 245 using ITK-SNAP software (version 3.4.0, www.itksnap.org, RRID: SCR 002010) 246 (Yushkevich et al., 2006). The resulting binary maps were then used to reconstruct individual 247 3D meshes of the cortical surfaces and aligned using a moving target-group average approach 248 based on curvature information (i.e., cortex-based alignment – CBA) to obtain an 249 anatomically aligned group-averaged 3D surface representation of all the subjects (Goebel et 250 al., 2006; Frost and Goebel, 2012). Functional data were analyzed (see Multivariate analysis 251 below) in the volume space and then projected to the average surface, to perform group 252 statistics and visualization in the aligned CBA space.

253 Univariate analysis

254 Single subject GLM analysis was performed on fMRI signal time courses normalized with 255 percentage transform in volume space. Next, they were mapped onto surface space by 256 sampling the values located between 1 mm below the gray/white matter boundary and up to 3 257 mm into the grav matter towards the pial surface using trilinear interpolation and averaging. 258 This sampling resulted in a single value per vertex in the subject's cortex mesh, and then the 259 values were aligned to the cortical group surface mesh using CBA. Random-effects GLM 260 analysis was performed on these individual time course data. The conditions were collapsed 261 across the speaker and vowel dimensions, thus yielding 12 predictors, 6 predictors for each 262 consonant times 2 tasks. Each predictor was convolved with a canonical double gamma 263 hemodynamic response function (HRF). Functional maps (i.e., average β values) were 264 calculated to assess sound-evoked fMRI responses during attend to vowels and attend to 265 consonants tasks (i.e., all sounds attend to vowels task > baseline; all sounds attend to

267 attend to vowels task activity (i.e., attend to consonants task > attend to vowels task). 268 Univariate stimulus effects were analyzed for each place of articulation feature independently 269 $(e.g., p_Con + f_Con + m_Con + p_Vow + f_Vow + m_Vow > t_Con + s_Con + n_Con + n_Con$ $t_Vow + s_Vow + n_Vow$). All functional contrast maps were corrected for multiple 270 271 comparisons by applying a permutation-based cluster-size threshold with an initial cluster 272 forming threshold of p = 0.05. The cluster-size threshold was based on the distribution of 273 maximum cluster sizes obtained in the 2,000 permutations, only considering clusters whose 274 size was larger than the 95% quantile.

consonants task > baseline). Task differences were analyzed for attend to consonants and

275 Multivariate analysis

266

276 To investigate the cortical representations of the different articulatory features, we used 277 MVPA in combination with a moving volumetric searchlight approach (Kriegeskorte et al., 278 2006). The purpose of the multivariate analysis was to decode articulatory features for each 279 syllable independent from their specific phonetic signatures or acoustical properties. 280 Therefore, we used a classification approach based on cross-decoding, generalizing place of 281 articulation features across different dimensions of manner of articulation. Specifically, we 282 trained a classifier to discriminate bilabial/labiodental vs. alveolar places of articulation using 283 syllables exhibiting two manners of articulation dimensions (e.g., stop and fricative) and 284 tested whether this training was transferable to syllables exhibiting a third (unseen) manner 285 dimension (e.g., nasal). MVPA generalization analysis was performed within each subject 286 using all three combinations of generalization (i.e., leaving one manner of articulation out for 287 testing at each split). The obtained averaged accuracies were submitted to a group analysis 288 using random permutations for significance and nonparametric permutation-based cluster-size 289 thresholding for multiple comparisons correction (see Statistical analysis – Group statistics 290 below).

291 Within-subject decoding

Before classification, BOLD time courses were selected by extracting the responses in each trial and each cortical voxel. The BOLD responses were estimated by fitting a standard HRF using multiple linear regression, taking the first three samples per trial after sound onset (i.e., corresponding to 6,000 ms). The regression coefficients resulting from the fitting of single trial data in the cortical mask were used to build an fMRI feature space (i.e., number of trials by number of cortical voxels), which was then used in the multivariate decoding. During the multivariate decoding, we kept trials belonging to each task separate.

299 We limited our analysis to gray matter voxels using a subject-specific cortical ribbon mask 300 based on intensity values in the T1-weighted images (i.e., cortical ribbon segmentation). We 301 constrained our analysis to the cortical ribbon, as we were mainly interested in cortical 302 processing. Ribbons allowed us to exclude volumes containing white matter and subcortical 303 voxels from the analysis, which increased the number of dimensions. For the searchlight 304 analysis, a sphere with radius of 5.5 voxels (i.e., 8.25 mm) was moved through the cortical 305 ribbon, which defined a feature space of 65,000 voxels on average. In each searchlight, we 306 performed generalization of place of articulation across manner of articulation using Linear 307 Discriminant Analysis (LDA) with diagonal covariance matrix. Given the large number of 308 searchlights, we employed a fast Matlab implementation described in Ontivero-Ortega et al. 309 (2017).

We also investigated the cortical representation of place of articulation independent of the task effect. We did so by performing classification and generalization across manner of articulation in the pooled data from the two tasks (i.e., attend to vowels and attend to consonants). By doing this we use twice the amount of data as in the previous analyses, and can expect more robust and reliable findings due to the power increase in single subject

analyses. To avoid any task bias, we controlled for the number of trials belonging to each task
in each cross-validation split maintaining a balanced contribution of each task during the
training and testing procedures. The decoding was performed as described above.

training and testing procedures. The decoding was performed as described above.

318 Multivariate statistical analysis – Group statistics

319 In each subject, classification accuracies were obtained, separately for each task (i.e., attention 320 to consonant and attention to vowel), for each searchlight centered on voxels belonging to the 321 subject's specific cortical ribbon. To perform a group analysis, these accuracies were 322 projected onto each subject's reconstructed cortical surface and then mapped back onto the 323 group average cortical sheet using CBA. After the re-alignment, we considered in the group 324 analysis only the vertices common to all subjects. To assess uncorrected significance of the 325 classifications at the group level, we considered a non-parametric permutation test at each 326 searchlight location independently. For each vertex, we took all the subject accuracies and the 327 associated mean, and, using a resampling approach, we estimated the probability that such 328 mean or a higher value could occur if the data were obtained under the null hypothesis (H0) 329 that the decoding population mean is at chance level (i.e., 50%). This probability is the p-330 value associated with the observed group mean. The resampling strategy was based on the 331 fact that, under H0, the likelihood of the observed accuracies is symmetric around chance 332 (i.e., 55% and 65% are equally likely if the population mean is 50%), and we can, therefore, 333 build many datasets by switching some of the subjects' accuracies around chance (Good, 334 2005, section 3.2.1). We ran 2,000 random permutations (Monte Carlo permutations) and for 335 each searchlight we determined the p-value as the ratio between how many times the mean of 336 a population resample equaled or exceeded the observed mean and the total number of 337 permutations (adding 1 to both numerator and denominator to avoid 0 p-values, hence the 338 lowest p-value was 1/2001).

339 Finally, to correct for multiple comparisons, we employed a permutation-based cluster-size 340 thresholding. We considered an initial Cluster Forming Threshold equal to 0.05 and, then, for 341 each permutation, we tagged as significant those vertices whose observed mean was higher 342 than the 1- quantile (obtained with the resampling procedure described above). For each of 343 these "false positive" maps, we determined the maximum extent of clusters (on the surface) 344 and built a distribution of cluster sizes across the 2,000 permutations. Clusters of significant 345 vertices, or equal to 0.05, in the observed data that were larger than the 95% quantile of the 346 obtained distribution were deemed significant and corrected for multiple comparisons. It is 347 worth mentioning that this procedure is based on a permutation strategy that does not suffer 348 from inflation of false positive rate, as recently shown in Eklund et al. (2016).

349 Results

350 Behavioral results in fMRI experiment

351 To test whether there were any significant difference in reaction times, a factorial analysis of 352 variance (ANOVA) was conducted, with task (attention to vowels and attention to 353 consonants), place of articulation (bilabial/labiodental and alveolar), manner of articulation 354 (stop, fricative, and nasal), speaker (speaker 1, speaker 2, and speaker 3), vowel (a, i, and u), 355 and subject (subject 1 to 14, treated as random factor) as main factors on correct trials only. 356 There was no effect across conditions in behavior. Reaction times for the vowel (mean = 843 357 ms, SD = 402) and the consonant task (mean = 820 ms, SD = 331) did not significantly differ 358 [F(1, 2541) = 1.26, p = 0.261]. There were no significant main effects or interactions (all p-359 values > 0.163).

360 Functional MRI univariate results during the attend to consonants and attend to vowels tasks

- 361 During both the attend to vowels and attend to consonants tasks, the stimuli evoked
- 362 significant BOLD responses in an extensive area of the superior temporal cortex,

encompassing early auditory areas (i.e., Heschl's gyrus and Heshl's sulcus), the planum
temporale, the superior temporal sulcus (STS) and gyrus (STG), and the posterior part of the
middle temporal gyrus (MTG). Additionally, the insula, supramarginal gyrus, intraparietal
sulcus and lateral prefrontal cortex including inferior frontal gyrus (IFG) were activated
during both tasks (Figure 2). Finally, univariate analyses did not reveal main effects of task or
main effect of place of articulation.

369 Functional MRI decoding results

To investigate the cortical representation of place of articulation features during the two
attentional conditions (i.e., attend to consonants and attend to vowels), we implemented a
classification method that relied on generalizing the discriminability of two places of
articulation features (i.e., bilabial/labiodental and alveolar) across variation of three manner of
articulation features (i.e., stop, fricative, and nasal).

375 Generalization maps after correction for multiple comparisons using cluster-size thresholding 376 (p-clust < 0.05) during the performance of the attend to vowels task are presented in Figure 377 3A. The generalization maps revealed successful decoding of place of articulation within 378 regions of the brain's language network, bilaterally. In the left hemisphere, clusters were 379 distributed across different regions including posterior temporal, temporo-parietal, insular, 380 anterior infero-frontal, frontal, and premotor medial regions. Specific regions included the 381 posterior superior temporal sulcus (STS), supramarginal gyrus (SMG), temporoparietal 382 junction, supplementary motor area (SMA), anterior insula, and anterior portion of the inferior 383 frontal gyrus (aIFG – pars orbitalis and pars triangularis). In the right hemisphere, clusters 384 included superior temporal, insular, inferior-motor and inferior-frontal regions. Specific 385 regions encompassed mid-posterior superior temporal plane (mSTG and pSTG), including 386 Heschl's Gyrus (HG), inferior central sulcus, subcentral gyrus, anterior insula, and posterior

(pars opercularis) and anterior (pars triangularis and orbitalis) parts of the inferior frontalgyrus (IFG).

Generalization maps for place of articulation during the performance of the attend to
consonants task revealed significant clusters in motor regions, insula, and anterior frontal
regions in the right hemisphere only (see Figure 3B). More specifically, these clusters
included the anterior-superior portion of the angular gyrus (AG), inferior precentral gyrus,
inferior central sulcus, subcentral gyrus, anterior-superior insula, posterior IFG, superior
frontal sulcus, and anterior middle frontal gyrus.

To test for the modulation of place of articulation feature representations between tasks, we compared the classification accuracies of these representations within each searchlight across the two tasks using a two-sample permutation test. This analysis yielded no significant clusters with correction for cluster-size in any of the hemispheres.

399 To explore possible tendencies in the data beyond the rigorous cluster-size thresholding 400 correction, we examined the maps without correction (i.e., p < 0.05 uncorrected). We 401 observed that the left hemisphere had more information (clusters) than the right hemisphere. 402 Particularly, we found a large continuous cluster of 134 square millimeters located in the left 403 anterior insula. For this cluster, the classification of place of articulation features during the 404 attend to vowels task exhibited higher accuracies than the classification of place of 405 articulation features during the attend to consonants task (see white outline in Figure 3 and 406 Figure 4). Aditionally, the posterior IFG bilaterally tended to show larger classification during 407 attend to vowels than during attend to consonants task (see Figure 4).

To increase the power in decoding at the single subject level, we renounced to task specificity by decoding the representation of place of articulation independently of the task effects. We

410 pooled together all the trials from the attend to vowels and attend to consonants tasks and 411 performed the classification of place of articulation ignoring whether a trial belonged to the 412 attend to vowels or attend to consonants task. Using twice as many data as in the task-specific 413 analysis, we expected this classification to be more robust and hence yield a higher number of 414 significant clusters. We found bilateral clusters in the angular, supramarginal, and inferior 415 frontal (i.e., pars opercularis and triangularis) gyri, the anterior insula, middle frontal rostral 416 areas, middle portion of the superior frontal and middle frontocaudal areas, and the 417 intraparietal sulcus. We also found clusters in the right posterior portion of the middle 418 temporal gyrus, the left precentral gyrus, and the right subcentral gyrus (see Figure 5).

419 Discussion

420 The present study investigated the spatial representation of place of articulation features 421 during the attentive perception of spoken consonant-vowel syllables (i.e., attend to vowels and 422 attend to consonants). Using identical acoustic input and generalization across variation due to 423 manner of articulation, we provided evidence for cortical representation of place of 424 articulation features. The generalization across manner of articulation allowed us to maximize 425 the acoustic invariance in fMRI classification (Correia et al., 2015) to counteract the 426 differences in acoustic properties (e.g., the variance of the second formant frequency of the 427 vowel segment due to perseverative coarticulation) that often accompany those in place of 428 articulation (Ladefoged and Johnson, 2010). We found representation of place of articulation 429 in separate tasks as well as in the pooled set (i.e., with both attention tasks). During the attend 430 to vowels task, generalization maps of place of articulation features indicated feature 431 sensitivity within bilateral clusters in superior and posterior temporal, insular, and prefrontal 432 regions. During the attend to consonants task, place of articulation features were represented 433 in temporoparietal, insular, and frontal regions within the right hemisphere only. The 434 representation of place of articulation features independent of task effects (i.e., performing the

generalization analysis in the pooled data) showed similar clusters to those obtained with each
task separately. In addition, this analysis yielded bilateral effects as well as a more prominent
contribution of frontal regions.

We observed that the brain represents the different features of place of articulation (i.e.,
bilabial/labiodental and alveolar) of speech sounds during attentive listening, as shown by
generalization across manner of articulation. This observation supports the relevance for place
in the context of sensorimotor representation of speech perception (Pulvermuller et al., 2006a;
Correia et al., 2015). The present result further supports previous findings that report withinplace feature differentiations in STS without generalization across manner (Davis and
Johnsrude, 2003; Scott and Johnsrude, 2003; Eulitz and Lahiri, 2004).

445 From a descriptive point of view, the cortical representation included temporal and frontal 446 regions during the attend to vowels task, and right middle and inferior frontal regions during 447 the attend to consonants task. This observation can be discussed in the light of differences in 448 (a) phonological feature specification for vowels and consonants, (b) stimulus processing for 449 consonants and vowels over time, and (c) attentional processing across the stimuli. 450 Consonants belonged to a *specific* feature class (i.e., bilabial/labidental or alveolar), whereas 451 vowels were equally represented across *all* syllables and place of articulation classes, see 452 Figure 1A (Eulitz and Lahiri, 2004; Scharinger et al., 2016). The temporal processing of 453 consonants and vowels in a CV structure could also have influenced the spatial distribution of 454 the cortical representation; it has been long recognized that the syllable onset has the most 455 relevant status in speech comprehension, whereas stimulus endings can be ignored more 456 easily (Marslen-Wilson and Zwitserlood, 1989).

The observed cortical representation could also have related to attentional processing and
variation of relevance-based selection across the two tasks. Attending to vowels meant

459	expecting the target to occur always during the <i>last</i> part of the syllable. However, although the
460	information related to the consonants is irrelevant for the task, preceding (irrelevant) linguistic
461	information cannot be ignored, as has been shown by studies on subliminal priming (Schütz et
462	al., 2007) and phonological priming (Bles and Jansma, 2008). Therefore, consonant-related
463	information could have still been linguistically processed during attention to vowels, thereby
464	impacting the respective representations within the language system. Attending to consonants
465	meant, in turn, expecting the target to occur always during the <i>first</i> part of the syllable, so the
466	information presented after the target (i.e., the vowel) could be easily ignored or discarded.
467	Consequently, the cortical representation of place of articulation features during the attend to
468	vowels task was found to include all areas initially expected for the attend to consonants task
469	(e.g., Correia et al., 2015). Similarly, consonant to vowel transition features may have been
470	amplified during the attend to vowels task, involving specific bilateral temporal regions
471	(Humphries et al., 2014). As during the attend to consonants task no transition (to a
472	subsequent vowel) should be amplified, brain regions dealing with such transitions were not
473	engaged. Future studies could also include other sequences of syllable components (e.g.,
474	Vowel-Consonant syllables or Consonant-Vowel-Consonant structures) to clarify the effect of
475	the relevance-based selection.

An alternative possibility pertains to general-feature sensitivity of the language system during attention to vowels and specific-feature selectivity during attention to consonants, as both syllable components can exhibit acoustic and articulatory properties (Ladefoged and Johnson, 2010). Given their relatively clearer acoustic features (Narayan et al., 2010), vowels could have had relatively higher perceptual (i.e., auditory) saliency in comparison with consonants, thus explaining the extended representation of place of articulation features during the attention to vowels task. However, this possibility is unlikely because our participants did not

differ in reaction times between tasks, and higher saliency should result in faster reactiontimes during detection and match to sample tasks.

485 The comparison between the two classification maps of place of articulation features yielded a 486 large continuous cluster (uncorrected) in the left anterior insula for attention to vowels vs. 487 attention to consonants. This result points towards a role of the anterior insula in the 488 modulation of the representation of place of articulation, determined by selective attention to 489 vowels in CV syllables. A relevance-based selection, resulting from the specificity of our 490 design, might have dictated a task set where attention to vowels required relatively greater 491 demands on top-down selectivity (i.e., in the attempt to ignore the first part of the syllable). 492 This interpretation is in agreement with the role of the anterior insula as part of a task-set 493 system (Dosenbach et al., 2006) in volitional top-down control (Nelson et al., 2010) and 494 alertness (Sadaghiani et al., 2010; Clemens et al., 2011). Moreover, the anterior insula has 495 been incorporated within the hierarchy of the language network (Davis and Johnsrude, 2003), 496 for example with a role in articulatory planning (Dronkers, 1996; Baldo et al., 2011).

497 The lack of significant clusters representing place of articulation features in other parts of the 498 temporal lobe (e.g., primary auditory cortex) was expected given our analysis approach. The 499 main purpose of combining MVPA with generalization was to maximize the extraction of the 500 information about the abstract (higher order) features under study (i.e., place of articulation) while minimizing the impact of variation in acoustic (i.e., manner of articulation) information. 501 502 Thus, the clusters show where information about place of articulation is represented, free from 503 perceptual properties and information processing of the sounds. Moreover, the univariate 504 analysis results support a critical role of the auditory cortices in speech perception and 505 auditory processing (see Figure 2). However, it should be noted that the univariate maps are 506 showing the main effect to sound compared to silence (i.e., overall relative changes of 507 activation in cortical regions in response to sound). Therefore, these maps reflect sound

responsive areas (to speech in our experiment), rather than isolated language-specific areas. The representation of place of articulation revealed by the generalization analysis in separate tasks was also confirmed when the trials from the two tasks were pooled. The additional clusters in frontal areas could have resulted from both the increase in the statistical power (i.e., using twice the number of trials in comparison to the task-specific analyses) and the higher sensitivity of multivariate pattern analysis (i.e, compared to univariate analysis; (Kriegeskorte et al., 2006; Norman et al., 2006; Mur et al., 2008).

515 Overall, our generalization approach showed a network of distributed brain regions (including 516 frontal and sensorimotor areas), where articulatory features are represented during the 517 perception, attention, and short-term memory storage within the delay-match-to-sample tasks. 518 This finding fits well with the cortical areas critical for language comprehension (e.g., the 519 angular and supramarginal gyri; Turken and Dronkers, 2011), sensorimotor integration 520 (Pulvermuller et al., 2006b; Hickok et al., 2011; Schomers and Pulvermüller, 2016), and 521 semantic knowledge retrieval (e.g., angular gyrus; Dronkers et al., 2004; Binder et al., 2009). 522 Moreover, lesion studies showing sentence comprehension impairment have also suggested a 523 role for dorsolateral frontal areas in auditory information integration (e.g., short-term verbal 524 memory rehearsal during sentence comprehension: Smith and Jonides, 1999; and the 525 representation of speech sequences: Fuster, 2001).

526 Considering the observed representation of place of articulation, it would seem reasonable to 527 analyze the individual phonological features in an independent manner. However, we could 528 not do so within our dataset due to the setup of the current experiment, in which the total 529 number of trials was limited. Limitations were related to the sparse sampling acquisition 530 necessary to avoid scanner noise interference, which increased the trial duration. In addition, 531 stimuli tokens were reduced to a subset required to introduce enough variance for 532 classification. These choices, though necessary for the purpose of the study, resulted in

selection of material with few individual place of articulation features. Previous studies have
also shown representation of individual abstract phonological features (Obleser et al., 2003,
2004), notions of underspecification (Scharinger et al., 2016), as well as a multidimensional
space of features (i.e., spectral peak or voice onset time) in the encoding of acoustic

537 parameters of speech (Mesgarani et al., 2008, 2014).

538 In summary, here we showed classification of place of articulation features generalized across

539 manner of articulation for the first time. Our results provide evidence for cortical

540 representation of place of articulation features during attentive syllable perception (i.e.,

541 attention to the different syllable components) in sensorimotor, temporoparietal, and frontal

542 areas within the language network. This cortical representation was revealed in additional

543 clusters (e.g., bilateral frontal and sensorimotor areas) when increasing the statistical power.

Additionally, we observed a more bilateral distribution for attend to vowels, a more unilateral

545 distribution for attend to consonants, and a trend towards a modulatory role of the left anterior

546 insula in selective attention of speech sounds. To conclude, these data support sensorimotor

547 integration during attentive speech perception and demonstrate that a generalization approach

can be used to exclude 'common factors', such as perceptual properties, from the analysis.

549 **References**

550	Arsenault JS, Buchsbaum BR (2015) Distributed Neural Representations of Phonological
551	Features during Speech Perception. J Neurosci 35:634–642 Available at:
552	http://www.jneurosci.org/cgi/doi/10.1523/JNEUROSCI.2454-14.2015.
553	Ashburner J, Friston KJ (2005) Unified segmentation. Neuroimage 26:839851 Available at:
554	http://dx.doi.org/10.1016%252?j.neuroimage.2005.02.018.
555 556	Baldauf D, Desimone R (2014) Neural Mechanisms of Object-Based Attention. Science (80-) 344:424–427 Available at: http://www.sciencemag.org/cgi/doi/10.1126/science.1247003.
557 558 559	Baldo J V, Wilkins DP, Ogar J, Willock S, Dronkers NF (2011) Role of the precentral gyrus of the insula in complex articulation. Cortex 47:800–807 Available at: http://dx.doi.org/10.1016/j.cortex.2010.07.001.
560 561 562	Binder JR, Desai RH, Graves WW, Conant LL (2009) Where is the semantic system? A critical review and meta-analysis of 120 functional neuroimaging studies. Cereb Cortex 19:2767–2796.
563 564 565	Bles M, Jansma BM (2008) Phonological processing of ignored distractor pictures, an fMRI investigation. BMC Neurosci 9:20 Available at: http://bmcneurosci.biomedcentral.com/articles/10.1186/1471-2202-9-20.
566	Bonte M, Hausfeld L, Scharke W, Valente G, Formisano E (2014) Task-Dependent Decoding
567	of Speaker and Vowel Identity from Auditory Cortical Response Patterns. J Neurosci
568	34:4548–4557 Available at:
569	http://www.jneurosci.org/cgi/doi/10.1523/JNEUROSCI.4339-13.2014.
570	Clemens B, Zvyagintsev M, Sack AT, Sack A, Heinecke A, Willmes K, Sturm W (2011)
571	Revealing the functional neuroanatomy of intrinsic alertness using fMRI: methodological
572	peculiarities. PLoS One 6:e25453 Available at:
573	http://www.ncbi.nlm.nih.gov/pubmed/21984928.
574	Correia JM, Jansma BM, Bonte M (2015) Decoding Articulatory Features from fMRI
575	Responses in Dorsal Speech Regions. J Neurosci 35:15015–15025 Available at:
576	http://www.jneurosci.org/cgi/doi/10.1523/JNEUROSCI.0977-15.2015.
577 578 579	Dale AM, Fischl B, Sereno MI (1999) Cortical surface-based analysis. I. Segmentation and surface reconstruction. Neuroimage 9:179–194 Available at: http://www.ncbi.nlm.nih.gov/pubmed/9931268.
580	Davis MH, Johnsrude IS (2003) Hierarchical processing in spoken language comprehension. J
581	Neurosci 23:3423–3431.
582 583	Di Salle F, Esposito F, Scarabino T, Formisano E, Marciano E, Saulino C, Cirillo S, Elefante R, Scheffler K, Seifritz E (2003) fMRI of the auditory system: understanding the neural

584 585	basis of auditory gestalt. Magn Reson Imaging 21:1213–1224 Available at: http://www.ncbi.nlm.nih.gov/pubmed/14725929.
586 587 588 589	Dosenbach NUF, Visscher KM, Palmer ED, Miezin FM, Wenger KK, Kang HC, Burgund ED, Grimes AL, Schlaggar BL, Petersen SE (2006) A core system for the implementation of task sets. Neuron 50:799–812 Available at: http://www.ncbi.nlm.nih.gov/pubmed/16731517.
590 591 592	Downer JD, Niwa M, Sutter ML (2015) Task Engagement Selectively Modulates Neural Correlations in Primary Auditory Cortex. J Neurosci 35:7565–7574 Available at: http://www.jneurosci.org/cgi/doi/10.1523/JNEUROSCI.4094-14.2015.
593 594	Dronkers NF (1996) A new brain region for coordinating speech articulation. Nature 384:159–161 Available at: http://www.ncbi.nlm.nih.gov/pubmed/8906789.
595 596	Dronkers NF, Wilkins DP, Van Valin RD, Redfern BB, Jaeger JJ (2004) Lesion analysis of the brain areas involved in language comprehension. Cognition 92:145–177.
597 598 599	Eklund A, Nichols TE, Knutsson H (2016) Cluster failure: Why fMRI inferences for spatial extent have inflated false-positive rates. Proc Natl Acad Sci 113:7900–7905 Available at: http://arxiv.org/abs/1606.08199.
600 601	Eulitz C, Lahiri A (2004) Neurobiological evidence for abstract phonological representations in the mental lexicon during speech recognition. J Cogn Neurosci 16:577–583.
602 603 604	Evans S, Davis MH (2015) Hierarchical organization of auditory and motor representations in speech perception: Evidence from searchlight similarity analysis. Cereb Cortex 25:4772–4788.
605 606 607	Evans S, McGettigan C, Agnew ZK, Rosen S, Scott SK (2016) Getting the Cocktail Party Started: Masking Effects in Speech Perception. J Cogn Neurosci 28:483–500 Available at: http://www.ncbi.nlm.nih.gov/pubmed/23647519.
608 609 610	Friederici AD, Singer W (2015) Grounding language processing on basic neurophysiological principles. Trends Cogn Sci 19:329–338 Available at: http://dx.doi.org/10.1016/j.tics.2015.03.012.
611 612 613	Fritz J, Shamma S, Elhilali M, Klein D (2003) Rapid task-related plasticity of spectrotemporal receptive fields in primary auditory cortex. Nat Neurosci 6:1216–1223 Available at: http://www.jneurosci.org/cgi/doi/10.1523/JNEUROSCI.1671-15.2015.
614 615 616	Frost M, Goebel R (2012) Measuring structural-functional correspondence: Spatial variability of specialised brain regions after macro-anatomical alignment. Neuroimage 59:1369–1381.
617 618	Fuster JM (2001) The prefrontal cortexan update: time is of the essence. Neuron 30:319–333 Available at: http://www.ncbi.nlm.nih.gov/pubmed/11394996.

619 620 621	Giraud A-L, Poeppel D (2012) Cortical oscillations and speech processing: emerging computational principles and operations. Nat Neurosci 15:511–517 Available at: http://dx.doi.org/10.1038/nn.3063.
622 623 624 625 626	Goebel R, Esposito F, Formisano E (2006) Analysis of functional image analysis contest (FIAC) data with brainvoyager QX: From single-subject to cortically aligned group general linear model analysis and self-organizing group independent component analysis. Hum Brain Mapp 27:392–401 Available at: http://doi.wiley.com/10.1002/hbm.20249.
627 628	Good PI (2005) Permutation, Parametric and Bootstrap Tests of Hypotheses. New York: Springer-Verlag. Available at: http://link.springer.com/10.1007/b138696.
629 630 631	Hickok G, Houde J, Rong F (2011) Sensorimotor Integration in Speech Processing: Computational Basis and Neural Organization. Neuron 69:407–422 Available at: http://dx.doi.org/10.1016/j.neuron.2011.01.019.
632 633	Hickok G, Poeppel D (2007) The cortical organization of speech processing. Nat Rev Neurosci 8:393–402 Available at: http://www.nature.com/doifinder/10.1038/nrn2113.
634 635	Humphries C, Sabri M, Lewis K, Liebenthal E (2014) Hierarchical organization of speech perception in human auditory cortex. Front Neurosci 8:1–12.
636 637	Hyafil A, Fontolan L, Kabdebon C, Gutkin B, Giraud AL (2015) Speech encoding by coupled cortical theta and gamma oscillations. Elife 4:1–45.
638 639 640 641	Jean T, Tournoux P (1988) Co-Planar Stereotaxic Atlas of the Human Brain: 3-D Proportional System: An Approach to Cerebral Imaging (Thieme Classics): J. Talairach: 9780865772939: Amazon.com: Books. Thieme:192 Available at: http://www.amazon.com/Co-Planar-Stereotaxic-Atlas-Human-Brain/dp/0865772932.
642 643 644	Kriegeskorte N, Goebel R, Bandettini P (2006) Information-based functional brain mapping. Proc Natl Acad Sci U S A 103:3863–3868 Available at: http://www.pnas.org/content/103/10/3863.abstract.
645	Ladefoged P, Johnson K (2010) A course in phonetics, 6th ed. Boston: Cengage Learning.
646 647	Lakatos P, Karmos G, Mehta AD, Ulbert I, Schroeder CE (2008) Entrainment of neuronal oscillations as a mechanism of attentional selection. Science 320:110–113.
648 649 650 651	Marques JP, Kober T, Krueger G, van der Zwaag W, Van de Moortele PF, Gruetter R (2010) MP2RAGE, a self bias-field corrected sequence for improved segmentation and T1- mapping at high field. Neuroimage 49:1271–1281 Available at: http://dx.doi.org/10.1016/j.neuroimage.2009.10.002.
652 653 654	Marslen-Wilson W, Zwitserlood P (1989) Accessing spoken words: The importance of word onsets. J Exp Psychol Hum Percept Perform 15:576–585 Available at: http://doi.apa.org/getdoi.cfm?doi=10.1037/0096-1523.15.3.576.

655 656 657	Mesgarani N, Cheung C, Johnson K, Chang EF (2014) Phonetic Feature Encoding in Human Superior Temporal Gyrus. Science (80-) 343:1006–1010 Available at: http://www.sciencemag.org/cgi/doi/10.1126/science.1245994.
658 659 660	Mesgarani N, David S V., Fritz JB, Shamma SA (2008) Phoneme representation and classification in primary auditory cortex. J Acoust Soc Am 123:899 Available at: http://scitation.aip.org/content/asa/journal/jasa/123/2/10.1121/1.2816572.
661 662 663 664	Moeller S, Yacoub E, Olman CA, Auerbach E, Strupp J, Harel N, Uğurbil K (2010) Multiband multislice GE-EPI at 7 tesla, with 16-fold acceleration using partial parallel imaging with application to high spatial and temporal whole-brain FMRI. Magn Reson Med 63:1144–1153.
665 666 667	Mur M, Bandettini PA, Kriegeskorte N (2008) Revealing representational content with pattern-information fMRIan introductory guide. Soc Cogn Affect Neurosci 4:101–109 Available at: http://scan.oxfordjournals.org/cgi/doi/10.1093/scan/nsn044.
668 669 670	Narayan CR, Werker JF, Beddor PS (2010) The interaction between acoustic salience and language experience in developmental speech perception: Evidence from nasal place discrimination. Dev Sci 13:407–420.
671 672 673 674 675	Nelson SM, Dosenbach NUF, Cohen AL, Wheeler ME, Schlaggar BL, Petersen SE (2010) Role of the anterior insula in task-level control and focal attention. Brain Struct Funct 214:669–680 Available at: http://www.pubmedcentral.nih.gov/articlerender.fcgi?artid=2886908&tool=pmcentrez&r endertype=abstract.
676 677 678 679	Norman KA, Polyn SM, Detre GJ, Haxby J V (2006) Beyond mind-reading: multi-voxel pattern analysis of fMRI data. Trends Cogn Sci 10:424–430 Available at: http://www.ncbi.nlm.nih.gov/entrez/query.fcgi?cmd=Retrieve&db=PubMed&dopt=Citat ion&list_uids=16899397.
680 681 682	Obleser J, Eulitz C, Reetz H, Lahiri A (2003) Phonological features of speech segments are reflected in the Auditory Evoked Brain Response around 100 ms post stimulus onset. Brain:1643–1646.
683 684 685	Obleser J, Lahiri A, Eulitz C (2004) Magnetic Brain Response Mirrors Extraction of Phonological Features from Spoken Vowels. J Cogn Neurosci 16:31–39 Available at: http://www.mitpressjournals.org/doi/10.1162/089892904322755539.
686 687 688	Ontivero-Ortega M, Lage-Castellanos A, Valente G, Goebel R, Valdes-Sosa M (2017) Fast Gaussian Naïve Bayes for searchlight classification analysis. Neuroimage:1–9 Available at: http://linkinghub.elsevier.com/retrieve/pii/S1053811917307371.
689 690 691	Polimeni JR, Fischl B, Greve DN, Wald LL (2010) Laminar analysis of 7T BOLD using an imposed spatial activation pattern in human V1. Neuroimage 52:1334–1346 Available at: http://dx.doi.org/10.1016/j.neuroimage.2010.05.005.

692	Pulvermuller F, Huss M, Kherif F, Moscoso del Prado Martin F, Hauk O, Shtyrov Y (2006a)
693	Motor cortex maps articulatory features of speech sounds. Proc Natl Acad Sci 103:7865–
694	7870 Available at: http://www.pnas.org/cgi/doi/10.1073/pnas.0509989103.
695	Pulvermuller F, Huss M, Kherif F, Moscoso del Prado Martin F, Hauk O, Shtyrov Y (2006b)
696	Motor cortex maps articulatory features of speech sounds. Proc Natl Acad Sci 103:7865–
697	7870 Available at: http://www.pnas.org/cgi/doi/10.1073/pnas.0509989103.
698	Sadaghiani S, Scheeringa R, Lehongre K, Morillon B, Giraud A-L, Kleinschmidt A (2010)
699	Intrinsic connectivity networks, alpha oscillations, and tonic alertness: a simultaneous
700	electroencephalography/functional magnetic resonance imaging study. J Neurosci
701	30:10243–10250 Available at:
702	http://www.jneurosci.org/cgi/doi/10.1523/JNEUROSCI.1004-10.2010.
703	Scharinger M, Domahs U, Klein E, Domahs F (2016) Mental representations of vowel
704	features asymmetrically modulate activity in superior temporal sulcus. Brain Lang
705	163:42–49 Available at: http://dx.doi.org/10.1016/j.bandl.2016.09.002.
706	Schomers MR, Pulvermüller F (2016) Is the Sensorimotor Cortex Relevant for Speech
707	Perception and Understanding? An Integrative Review. Front Hum Neurosci 10
708	Available at: http://journal.frontiersin.org/Article/10.3389/fnhum.2016.00435/abstract.
709 710 711	Schütz K, Schendzielarz I, Zwitserlood P, Vorberg D (2007) Nice wor_if you can get the wor_: Subliminal semantic and form priming in fragment completion. Conscious Cogn 16:520–532.
712 713 714	Scott SK, Johnsrude IS (2003) The neuroanatomical and functional organization of speech perception. Trends Neurosci 26:100–107 Available at: http://www.sciencedirect.com/science/article/pii/S0166223602000371.
715	Setsompop K, Gagoski BA, Polimeni JR, Witzel T, Wedeen VJ, Wald LL (2012) Blipped-
716	controlled aliasing in parallel imaging for simultaneous multislice echo planar imaging
717	with reduced g-factor penalty. Magn Reson Med 67:1210–1224.
718 719 720	Smith EE, Jonides J (1999) Storage and executive processes in the frontal lobes. Science 283:1657–1661 Available at: http://www.sciencedirect.com/science/article/pii/095943889390204C.
721	Smith SM, Jenkinson M, Woolrich MW, Beckmann CF, Behrens TEJ, Johansen-berg H,
722	Bannister PR, Luca M De, Drobnjak I, Flitney DE, Niazy RK, Saunders J, Vickers J,
723	Zhang Y, Stefano N De, Brady JM, Matthews PM (2004) Advances in Functional and
724	Structural MR Image Analysis and Implementation as FSL Technical Report TR04SS2.
725	Neuroimage 23(S1):208–219.
726	Turken AU, Dronkers NF (2011) The neural architecture of the language comprehension
727	network: converging evidence from lesion and connectivity analyses. Front Syst
728	Neurosci 5:1 Available at:
729	http://journal.frontiersin.org/article/10.3389/fnsys.2011.00001/abstract.

730	Vaughan JT, Garwood M, Collins CM, Liu W, Delabarre L, Adriany G, Andersen P, Merkle
731	H, Goebel R, Smith MB, Ugurbil K (2001) 7T vs. 4T: RF power, homogeneity, and
732	signal-to-noise comparison in head images. Magn Reson Med 46:24-30.

733 Yacou	ub E, Shmuel a, Pfeuffer J, Van De Moortele PF, Adriany G, Andersen P, Vaughan JT,
734 N	Merkle H, Ugurbil K, Hu X (2001) Imaging brain function in humans at 7 Tesla. Magn
735 F	Reson Med 45:588–594 Available at: http://www.ncbi.nlm.nih.gov/pubmed/11283986.

Yushkevich PA, Piven J, Hazlett HC, Smith RG, Ho S, Gee JC, Gerig G (2006) User-guided
3D active contour segmentation of anatomical structures: Significantly improved
efficiency and reliability. Neuroimage 31:1116–1128.

740 Legends

741 Figure 1.

742 Stimuli and experimental paradigm. A) Spoken stimuli matrix and articulatory properties. The 18 syllables were 743 selected according to the place of articulation (i.e. bilabial/labiodental and alveolar) and manner of articulation 744 (i.e. stop, fricative, and nasal). B) Experimental procedure and task description. Example of a typical match-to-745 sample trial during attend to vowels task. Subjects received instructions per block, in which they attended to 746 consonants or vowels, respectively, and carried out a match-to-sample decision within slow-event related trials. 747 Each block started with the visual presentation of a task cue (i.e., attention target vowel or consonant) indicating 748 which task to perform in the next 27 trials. Each trial started with a fixation period in which a fixation cross was 749 presented at the center of screen, together with a syllable sound (340 ms). After a jittered inter-stimulus interval 750 (ISI jitter: 6 - 8 seconds) a visual cue (i.e., a written letter, vowel or consonant) was presented for 2 seconds, 751 followed by the immediate subject's response by pressing a button either with the right index finger (for match 752 trials) or middle index finger (for mismatch trials). The response was followed by a jittered ISI (4 - 10 seconds)753 to complete the jittered intertrial interval period (ITI: 12-16 seconds) before the next trial started. C) Schematic 754 representation of the fMRI acquisition sequence and its relationship to the syllable sounds presented to the 755 participants. Vow: attentd to vowels task, Con: attend to consonants task, Syll: syllable, TA: time of acquisition, 756 TR: time for repetition, Phase: opposite phase encoding volumes acquired for distortion correction.

757 Figure 2.

758

Functional maps depicting the overall pattern of sound-evoked cortical responses during performance of A)
attend to vowels (i.e., all sounds attend to vowels > baseline) and B) attend to consonants task (i.e., all sounds

Brain areas sensitive to speech sound processing during attend to vowels and attend to consonants tasks.

- $761 \qquad \text{attend to consonants} > \text{baseline}). P-value at cluster threshold (p-clust) < 0.05. Maps are visualized on inflated on inflated on the second sec$
- 762 and aligned group-averaged representations of the left (LH) and right (RH) hemispheres of the fourteen subjects

763 (light gray: gyri and dark gray: sulci). Color scale indicates average beta values.

764 Figure 3.

765 Cortical representation of place of articulation features, generalized across manner of articulation. Generalization 766 maps depicting classification accuracies during A) attend to vowels and B) attend to consonants tasks. Insula 767 cluster represents the outline (i.e., while line) of the largest continuous uncorrected cluster from the differences between the two tasks. *P*-value at cluster-size threshold (pclust) < 0.05. Searchlight radius 8.25 mm. Maps are
visualized on inflated and aligned group-averaged representations of the left (LH) and right (RH) hemispheres of
the fourteen subjects (light gray: gyri and dark gray: sulci). Color scale indicates classification accuracy (Acc)
percentages.

772 Figure 4.

773 Differences in classification of place of articulation features, generalized across manner of articulation. Maps 774 depicting the difference in the classification accuracies between attend to vowels and attend to consonants tasks. 775 The largest continuous cluster (i.e., area 134 square millimiters) was found in the left anterior insula, outlined in 776 white. P-value uncorrected (puncorr) < 0.05. Searchlight radius 8.25 mm. Maps are visualized on inflated and 777 aligned group-averaged representations of the left (LH) and right (RH) hemispheres of the fourteen subjects 778 (light gray: gyri and dark gray: sulci). Yellow-orange clusters show larger classification accuracies during the 779 attend to vowels task and green-blue clusters show lager classification accuracies during attend to consonants 780 task. Please note that the color scale does not directly relates to the colors used in Figure 3. Color scale indicates 781 differences in the classification accuracy (Acc) percentages.

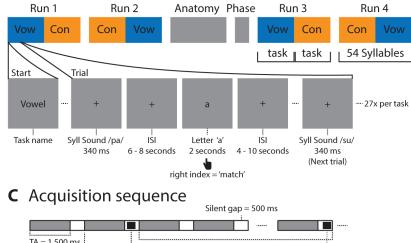
782 Figure 5

783 Cortical representation of place of articulation features, generalized across manner of articulation, pooling all 784 trails from two tasks. Generalization maps depicting classification accuracies calculated by pooling all trails 785 from attend to vowels and attend to consonants tasks. Corrected for multiple comparison with cluster-size 786 thresholding. Insula cluster represents the outline (i.e., white line) of the largest continuous uncorrected cluster 787 from the differences between the two tasks. P-value at cluster-size threshold (p_{clust}) < 0.05. Searchlight radius 788 8.25 mm. Maps are visualized on inflated and aligned group-averaged representations of the left (LH) and right 789 (RH) hemispheres of the fourteen subjects (light gray: gyri and dark gray: sulci). Color scale indicates 790 classification accuracy (Acc) percentages.

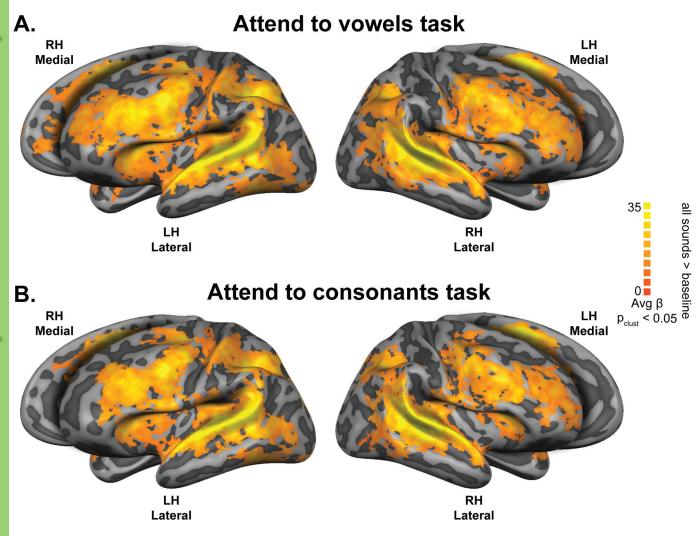
A Stimuli matrix

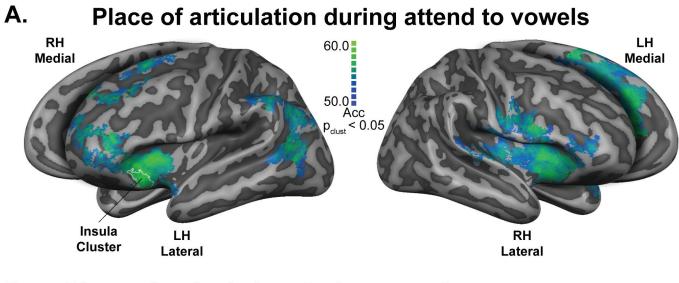
		Place of Articulation		
Manner of Articulation		Bilabial Labiodental	Alveolar	
	Stop	ʻp' /pa/ /pi/ /pu/	't' /ta/ /ti/ /tu/	
	Fricative	'f' /fa/ /fi/ /fu/	ʻs' /sa/ /si/ /su/	
	Nasal	'm' /ma/ /mi/ /mu/	ʻn' /na/ /ni/ /nu/	

B Experiment structure

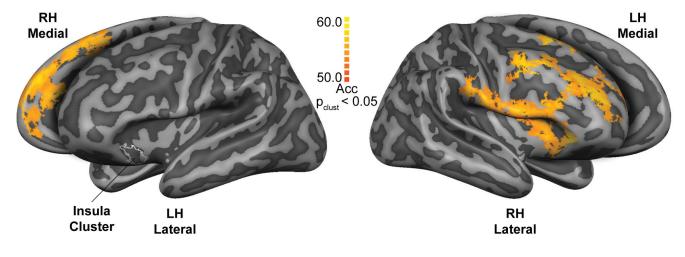


TA = 1,500 ms I TR = 2,000 ms	i Syll sound = 340 ms	ITI = 12 - 16 seconds	Syll sound = 340 ms (Next trial)
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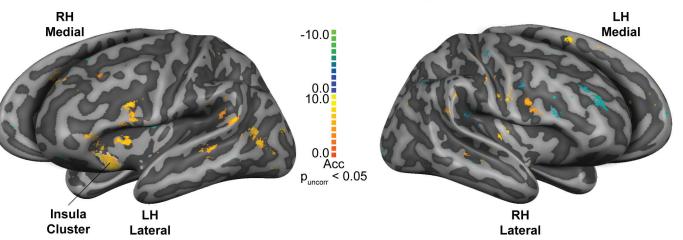




B. Place of articulation during attend to consonants



Differences in the classification of place of articulation



Place of articulation pooling all trials from both tasks

