Infants' preference for speech decomposed: Meta-analytic evidence

Cécile Issard¹, Sho Tsuji², & Alejandrina Cristia¹

¹ Laboratoire de Sciences Cognitives et Psycholinguistique, Ecole Normale Supérieure, Département d'Études Cognitives

² International Research Center for Neurointelligence, The University of Tokyo

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Correspondence concerning this article should be addressed to Cécile Issard,
Laboratoire de Sciences Cognitives et Psycholinguistique, Département d'Études Cognitives,
Ecole Normale Supérieure, 29 rue d'Ulm, 75005 Paris, France. E-mail:
cecile.issard@gmail.com

Abstract

Previous experimental studies suggested that infants show a preference for speech over auditory competitors, but the mechanism underlying the preference remains unclear, with some results pointing to a preference for (a) familiar sounds; (b) vocal sounds; or (c) natural sounds more generally, with potentially changes over development. To shed light on whether some of these mechanisms may be at play, we conducted a meta-analysis of experiments testing speech preference in infants. Infants reliably preferred speech over other sounds, but this preference was not significantly moderated by familiarity, vocal quality, or naturalness of the competitor. Also, we found no effect of age: infants showed the same strength of preference throughout the first year of life. Speech therefore appears to be preferred from birth, even to other natural or vocal sounds. These results contradict current views of the literature, and call for further investigation of the phenomenon.

Keywords: Meta-analysis, infants, speech preference, auditory development, natural sounds

Word count:

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Highlights

- Infants reliably prefer natural speech over other types of sounds, from birth to the end of the first year of life
- Speech is preferred over both artificial and other natural sounds
- Speech is preferred over both non-vocal and other vocal sounds
- The difference between whether infants are familiar or not with the language used was not significant

Introduction

Communication is a crucial behavior in humans, crucially supporting complex social interactions. Previous work has suggested communication skills develop from early infancy, manifesting in an early preference for speech over other types of sound (Ecklund-Flores & Turkewitz, 1996; Vouloumanos, Hauser, Werker, & Martin, 2010; Vouloumanos & Werker, 2007). This preference may emerge through a variety of simpler mechanisms, saliently including a preference for familiar over unfamiliar sounds, natural over artificial sounds, and vocal over non-vocal sounds (Figure 1). Here, we synthesize the available empirical data on infants' preferences for speech over non-speech sounds to assess the explanatory role of these mechanisms.

One mechanism could involve familiarity: Perhaps infants prefer speech to other sounds because speech is a frequent sound in their experience. Newborns prefer their native speech to prosodically distinct foreign speech (e.g., Mehler et al., 1988; Moon, Cooper, & Fifer, 1993), which supports a preference for sound patterns heard frequently in the womb. If this mechanism is at play in speech preference, infants should show a stronger preference for

speech over other sounds when tested in their native language, and a weaker preference when tested with a foreign language.

A second mechanism could involve a preference for natural over artificial sounds. Natural sounds are those produced by biological systems, such as heart beats, step sounds, or animal vocalizations, and environmental/geophysical sounds, such as wind, rain, or the sound of a river. Natural sounds are processed more efficiently by the auditory system, from the cochlea (Smith & Lewicki, 2006) to the auditory cortex (see Mizrahi, Shalev, & Nelken, 2014 for a review). If this mechanism (partially) explains speech preferences, then we should observe that the preference for speech over artificial competitors is stronger than the preference observed when contrasting speech against another natural sounds.

A third mechanism would rely on a preference for vocal sounds, which may even be shared with other non-human primates given the prevalence of oral communication across primate species. Vocal sounds have acoustic signatures, since they are characterized by modulations introduced by the vocal tract, with harmonically related energy peaks. If this mechanism explains at least partically speech preferences, the preference for speech over non-vocal competitors is stronger than the preference observed when contrasting speech against another vocal sounds.

Development may affect the preference for speech in various ways. For example, whereas newborns do not prefer speech over monkey calls, there is some evidence that three-month-olds do (Vouloumanos et al., 2010). More broadly, both maturation and experience could affect the extent to which naturalness, familiarity, and vocal quality affect infants' preferences as a function of age.

A meta-analytic approach

In sum, previous results on infants' preference for speech are broadly compatible with a preference for natural over artificial, vocal over non-vocal, and/or familiar over unfamiliar sounds, potentially interacting with infants' age. In this paper, we seek to directly test these potential mechanisms by employing a meta-analytic approach.

Meta-analyses are quantitative syntheses of experiments testing a comparable phenomenon with comparable approaches. In a nutshell, meta-analyses allow us to draw statistics over multiple experiments. This is important because any single experiment is a noisy window onto the underlying reality of a given phenomenon: Despite experimentalists' best efforts, a given experiment can at best tell us about what a specific group of participants, presented with a specific set of stimuli, in a specific point in time has done. As a result, even if we assume that all experiments that are carried out are systematically published, meta-analyses are useful because they allow us to integrate those individual and specific results into a larger picture. By integrating data across different laboratories, they provide evidence for the generalizability of effects across labs, and facilitate comparisons between experimental results.

Moreover, meta-analyses allow direct comparison and integration of data in a principled statistical approach. To begin with, data drawn from larger samples and with higher precision is given more weight than smaller and less precise estimates. By combining studies to obtain larger samples, meta-analyses can reveal small effects not obvious in individual experiments.

Additionally, a meta-analysis allows us to statistically test and/or statistically control for different factors. This allows us to integrate data from experiments that vary in their methodology, as well as test the effect of factors that are not part of the original design, by redescribing the stimuli used as a function of those factors. For instance, a study measuring

preference for native speech over native backward speech provides data on a natural versus artificial contrast, as well as a vocal versus non-vocal contrast.

We can also draw a developmental timeline across the age range covered by the literature, beyond age groups tested within papers. This is particularly useful for potential changes with age, as in the field of infant studies it is common to make age-related statements using the demonstrably problematic method of concluding that there is an interaction without actually testing for it statistically (e.g. Gelman & Stern, 2006). It is therefore important to directly test for interactions with an actual statistical approach. To give an example from a previous developmental meta-analysis, it had been proposed that infants' preference for novel or familiar items related to infants' age such that, all things equal, younger infants showed familiarity preferences whereas older infants exhibited novelty preferences (Hunter & Ames, 1988). However, Bergmann and Cristia (2016) found stable familiarity preferences for word segmentation in natural speech across the first two years; and Black and Bergmann (2017) found a stable novelty effect for artificial grammars implemented in synthesized speech, whereas those implemented in natural speech led to stable familiarity preferences. Meta-analyses are therefore important to statistically and systematically test the theoretical predictions proposed in qualitative reviews, and refine current readings of a literature.

Finally, meta-analyses offer tools to detect publication bias in the literature. By aggregating all the available evidence for a phenomenon, we can see if the distribution of effect sizes has an unexpected shape, typically with an excess of positive results due to the difficulty to publish null or negative results. We can further integrate this information, and derive a new estimate of the overall effect size.

The present study

Although we were primarily motivated by a theoretical quest, we know meta-analyses provide a unique vantage point on a body of work as a whole. We therefore first describe this body of literature, as this allows us to provide recommendations for further experimental work in cases where more evidence is necessary to find reliable answers.

Next, we check for how strong infants' preference for speech over other types of sounds is according to the public body of literature. We additionally assess this body of data for evidence of publication bias.

We then turn to our key interest, namely shedding light on the potential mechanisms underlying infants' speech preferences. Meta-regressions assess whether the proposed mechanisms of familiarity, naturalness, and vocal quality drive this preference, and how the preference develops over the first year of life. Assuming all three mechanisms are at play, and further assuming that the definition of the preferred stimulus narrows with age, we predicted that infants will show (see Figure 2):

- 1. a greater preference for native speech over non-speech as a function of age, but a smaller preference for foreign speech over non-speech with age;
- 2. a greater preference for speech over other natural sounds as a function of age, but a preference for speech over artificial sounds that is stable over development;
- 3. a greater preference for speech over other vocal sounds as a function of age, but a stable preference for speech over non-vocal sounds over development.

Methods

This meta-analysis was carried out following PRISMA recommendations (Moher, Liberati, Tetzlaff, Altman, & Group, 2009) so as to yield a meta-analysis of the highest level

of quality. In addition, we provide information on all steps (including PRISMA checklist, data, and code) for full transparency and accountability via online supplementary materials; https://osf.io/4stz9/?view_only=d0696591ebf34bfc8430f848cd945ca8.

Literature search

We composed the initial list of papers with suggestions by experts (authors of this work); one google scholar search (("speech preference" OR "own-species vocalization") AND infant - "infant-directed"), the same search in PubMed and PsycInfo (last searched on 2019-09-24); and a google alert. We also inspected the reference lists of all included papers. Finally, we emailed a major mailing list to ask for missing data. We received only two replies, one of which revealed a formerly undiscovered published study, but no unpublished data was made available to us.

Inclusion criteria

After a first screening based on titles and abstracts using more liberal inclusion criteria, we decided on final inclusion based on full paper reading. We included experiments that tested human infants from birth to one year of age, and contrasted speech sounds with any other type of sound, measuring behavioral preferences to the sounds (e.g., looking times). If a paper reported results from neurotypical and at-risk infants, we included only the data from the neurotypical group.

Given our key interest in the preference for speech over other sounds, we excluded studies that contrasted two different speech sounds (e.g., foreign vs. native language, or adult vs. child-directed speech, or mother vs. stranger's voice); or two different non-speech sounds (e.g., backward speech vs. animal vocalizations). In addition, we excluded experiments where it was unclear how to code the contrastst presented to the infants according to our three

mechanistic explanations. This meant the exclusion of two types: (1) experiments that intentionally mixed speech with other vocal sounds within the same sound condition were excluded (e.g., speech and emotional vocalizations vs. backward speech and emotional vocalizations); (2) experiments where speech was presented in the mother's voice (which thus counfounds between speech and individual voice recognition for our familiarity factor). Finally, we excluded neuroimaging experiments to avoid mixing results from different brain regions with different response profiles. We included published (i.e., journal articles) as well as unpublished works (i.e., doctoral dissertations) as long as sufficient information was provided.

A PRISMA flow chart summarizes the literature review and selection process (Figure 3). The full list of the papers that were inspected together final inclusion decisions are available in a decision spreadsheet (see the online supplementary materials; https://osf.io/4stz9/?view_only=d0696591ebf34bfc8430f848cd945ca8).

Coding

Data were coded by the first author. In addition, 20% of the papers were randomly selected to be coded by the last author independently, with disagreements resolved by discussion. There were 10 disagreements out of a total of 260 fields filled in, and they were indicative of the coders not following the codebook, which led to a revision of all data in four variables.

The critical variables for our purpose are key stimuli characteristics, infant age, and testing method (central fixation, high amplitude sucking, head-turn preference procedure). As for key stimuli characteristics, we coded familiarity, naturalness, and vocal quality, as follows.

For familiarity, we considered the language in which the speech sounds were recorded

(native or foreign).

For naturalness, the competitor sound was coded as natural if it was produced by a biological organism without any further acoustic manipulation. Natural competitors included animal calls, environmental sounds (e.g. wind or water sounds), heartbeat, bird song, non-speech vocalizations (e.g. laughters or coughs). If the authors applied acoustic manipulations, the competitor was coded as artificial. Artificial competitors included sine-wave speech, filtered speech¹, white noise, instrumental music, and speech with altered rhythmic structure. The only exception was for newborn experiments presenting low-pass filtered speech mimicking the filtering applied by the womb. Given the recency of the intra-uterine environement to newborns (about 2 days), we coded these as natural.

For vocal quality, the competitor sound was considered as vocal if it was produced by an animal vocal tract (human or not), either original or modified. Vocal competitors included non-speech vocalizations, animal calls, bird songs, and filtered speech. Non-vocal competitors included backward speech (that has abrupt closures that cannot be produced by the vocal tract), white-noise, environmental sounds, instrumental music, heartbeat, and sine-wave speech (that lacks the harmonic structure introduced by the natural resonance of the vocal tract).

We coded all the statistical information reported in the included papers. If reported, we coded the mean score and the standard deviation for speech, and the other sound separately. When infant-level data was provided, we recomputed the respective mean scores and standard deviations based on the reported individual scores. If reported, we also coded the t-statistic between the two sound conditions, or an F-statistic provided this was a

¹In the case of filtered speech, the modulations introduced by the vocal tract are still present at the retained frequencies, and formant transitions are consistent with vocal production constraints. For this reason, filtered speech can be considered as vocal but not natural. Because the womb acts as a low-pass filter, newborn infants are familiar with low-pass filtered speech, but this familiarity fades after birth.

two-way comparison. If effect sizes were directly reported as a Cohen's d or a Hedges' g, we also coded this.

Effect sizes

Once the data were coded, we extracted effect sizes, along with their respective variance. Effect sizes were standardized differences (Cohen's d) between response to speech vs. the competitor. If effect sizes were not directly reported in the papers, we computed them using the respective means and SDs (Lipsey & Wilson, 2001), or a t- or F-statistic (Dunlap, Cortina, Vaslow, & Burke, 1996). As our effect sizes came from within-subject comparisons (e.g., looking time of the same infant during speech and during monkey calls), we needed to take into account the correlation between the two measurements in effect sizes and effect size variances computations. We computed this correlation based on the t-statistic, the respective means and SDs (Lipsey & Wilson, 2001) if they were all reported; or imputed this correlation randomly if not. We finally calculated the variance of each effect size (Lipsey & Wilson, 2001). Cohen's d were transformed to Hedges' g by multiplying d by a correction for small sample sizes based on the degree of freedom (Borenstein, Hedges, Higgins, & Rothstein, 2011).

We did not center age because our hypotheses included a developmental progression from birth to the end of the first year of life. We were therefore interested in the intercept at age 0 (i.e., birth).

Analyses use the R (R Core Team, 2018) package Robumeta (Hedges, Tipton, & Johnson, 2010), which allows us to fit meta-analytic regressions that take into account the correlated structure of the data, when repeated measures are obtained from the same infant groups within papers.

Results

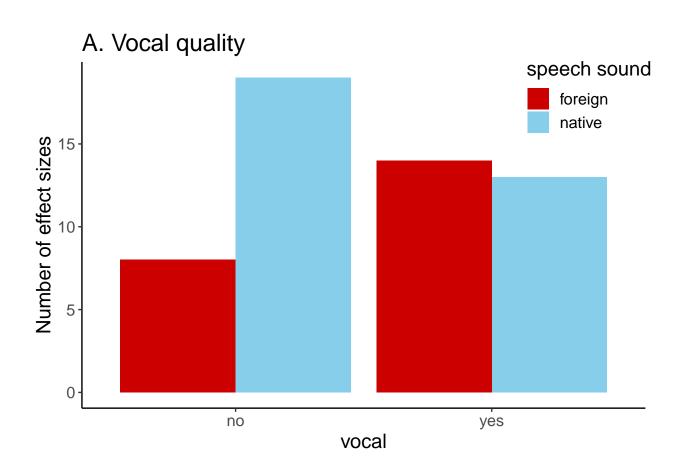
Database description

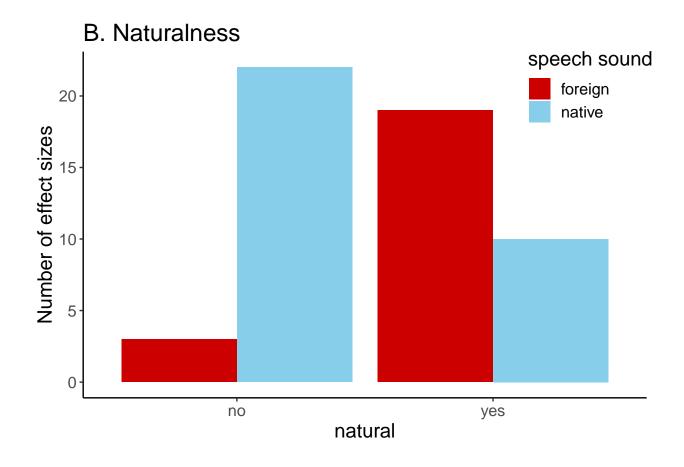
We found a total of 18 publications (labeled with an asterisk in the reference list) reporting 33 experiments, and 54 (not mutually independent) effect sizes, see Figure 5. 17 papers have been submitted to or published in peer-reviewed journals (Colombo & Bundy, 1981; Cooper & Aslin, 1994; Curtin & Vouloumanos, 2013; Ecklund-Flores & Turkewitz, 1996; Santolin, Russo, Calignano, Saffran, & Valenza, 2019; Segal & Kishon-Rabin, 2011; Shultz & Vouloumanos, 2010; Sorcinelli, Ference, Curtin, & Vouloumanos, 2019; Spence & DeCasper, 1987; Vanden Bosch der Nederlanden & Vouloumanos, 2020; Vouloumanos & Curtin, 2014; Vouloumanos, Druhen, Hauser, & Huizink, 2009; Vouloumanos et al., 2010, 2010; Vouloumanos & Werker, 2004, 2007; Yamashiro, Curtin, & Vouloumanos, 2019). The remaining 1 publication contributing 1 effect size was a thesis (Ference, 2018).

Experiments tended to have small sample sizes, with a median N of 16 children (Range = [60, 4], M = 19.83, Total: 727). Infants ranged from 0 to 12 months (1.50 to 380.50 days), although the majority were under 6 months of age (72.73% of the experiments). Individual samples comprised 46% of female participants on average. Infants were native of 6 different languages across the whole database (English, French, Russian, Yiddish, Hebrew, Italian). Experiments were performed in 10 different laboratories from 4 different countries (United States, Canada, Israel, Italy). 3 experimental methods were used: 25 experiments used Central Fixation (CF) (also called sequential looking preference procedure) (Colombo & Bundy, 1981; Cooper & Aslin, 1994; Curtin & Vouloumanos, 2013; Ference, 2018; Santolin et al., 2019; Segal & Kishon-Rabin, 2011; Shultz & Vouloumanos, 2010; Sorcinelli et al., 2019; Vanden Bosch der Nederlanden & Vouloumanos, 2020; Vouloumanos & Curtin, 2014; Vouloumanos et al., 2009, 2010; Vouloumanos & Werker, 2004; Yamashiro et al., 2019); 3 used High-Amplitude Sucking (HAS) (Spence & DeCasper, 1987; Vouloumanos et al., 2010;

Vouloumanos & Werker, 2007); and 5 used Head-turn Preference Procedure (HPP) (Ecklund-Flores & Turkewitz, 1996). Trial length was fixed in 8 experiments, and infant-controlled in 24 experiments.

Speech sounds were spoken by a female in 31 out of 33 experiments, with an infant-directed prosody in 16 out of the 33 experiments. Speech was presented in isolated segments (i.e. words or syllables) in 5 experiments, and full sentences or passages in 14 experiments. Speech stimuli were recorded in the infant native language in 60.61% of the experiments. Strinkingly, experiments using the infants' native language tested infants from 0 to 12 months of age, whereas experiments using a foreign language only tested infants from 3 to 9 months of age (see Figure 9). The competitor sound was vocal in 51.52% of the experiments. The competitor sound was natural 51.52% of the experiments. The stimuli characteristics are sumarized on Figures ?? and 4.





Average effect size

Integrating across all experiments in a meta-analytic regression without any moderator, we found an average effect size g of 0.34 (SE = 0.06, CI = [0.21, 0.47]) (Table 1, and Figure 5, diamond), corresponding to a medium effect size. Heterogeneity among effect sizes was estimated at $\tau^2 = 0.12$ (I² = 76.26%), which was significant (Q = 188.62, p<0.001) despite the removal of outliers before running the model. This strongly suggest differences across experiments, and invites analyses using regressors.

Publication bias

We checked for the presence of a potential publication bias in the body of literature by studying the relationship between standard errors of effect sizes as a function of Hedges' g (see funnel plot in Figure 6).² A regression test on these data was significant (z = 7.05, p = 0.00), as was the Kendall's tau rank correlation test for funnel plot asymmetry (Kendall's tau = 0.46, p = 0.00), consistent with a publication bias in the literature.

To check whether this bias fully explains infants' speech preference, we symmetrized the funnel plot with the "trim and fill" method (Duval & Tweedie, 2000). To symmetrize the funnel plot, 11 (SE = 4.71) missing experiments were needed on the left side of the plot. The corrected effect size was estimated at 0.23 (SE = 0.07) after filling in the 11 missing experiments, which is still significantly different from zero. Thus, even correcting for a potential publication bias, we still find statistical evidence for infants' preferring speech over competitors.

Moderator analyses

We then tested if heterogeneity could be counted for by the mechanistic explanations described in our introduction. Following our hypotheses, we fit a meta-analytic model with the following moderators:

- mean age of children;
- familiarity with the language used (native or foreign);
- naturalness of the contrastive sound (coded as yes if it was natural and no otherwise);

²If the literature is not biased, effect sizes should be evenly distributed around the mean effect size, with increasing standard error as they go away from the mean effect size (both in the positive and negative directions, white triangle in the funnel plot). This is reflected by a symmetrical funnel plot, with no linear relationship between effect sizes and standard errors.

• vocal quality of the contrastive sound (coded as yes if it was vocal and no otherwise).

Given the number of effect sizes available in the literature, these moderators were specified without interactions to avoid overfitting.

None of the moderators was significant (see Figures 7, 8, and 9; and Table 2).

We also tested each of our three hypotheses by three separate models for each moderator and its interaction with age. None of them yielded a significant effect or interaction (Table 3, 4, and 5).

Discussion

Our meta-analysis synthesizes the available literature on infants' preference for speech sounds. When all experiments were considered together with no moderators, we found a sizable intercept (g=0.34). For comparison, the average effect for native vowel discrimination using looking time methods is estimated at 0.25 (Tsuji & Cristia, 2014, data inspected in http://metalab.stanford.edu on 2019-10-18). Our meta-analysis shows that this preferential processing of speech sounds is observable from birth on. We had hypothesized age to play a major role, because it may correlate with a reshaping of the category definition for speech itself. Indeed, experiments comparing processing of human speech against human non-speech as well as animal vocalizations more generally (McDonald et al., 2019; Vouloumanos et al., 2010) often discuss these age-related differences in categorization of these sounds. Surprisingly, age did not significantly moderate the overall preference for speech, as shown by the null estimate of this moderator (Table 1). This result was replicated in a separate model for age only, which showed a significant intercept, similar to the intercept found in the meta-regression with no moderator (Supplementary results S5). Crucially, age was not centered. This intercept therefore provides an estimate of the effect size at age 0, i.e. at birth. Moreover, the scatterplot of effect sizes as a function of age reveals clearly no change with

age even when plotted without other moderators (Supplementary figure S6). This null effect of age, combined with the sizeable intercept, confirms that infants reliably prefer speech over other types of sounds from birth.

The significant heterogeneity we found among the literature suggests that underlying factors modulate this effect. We had predicted infants' speech preference to be larger when the competitor was an artificial sound than when it was a natural one; when the competitor was non-vocal; and when the speech was in the infants' native language. In fact, we were unable to disprove the null hypothesis of no difference for all three factors. Distributions of effect sizes for experiments varying along the three dimensions widely overlap, and the confidence intervals of all conditions overlap almost exactly (Figures 7, 8, and 9). Moreover, Table 1 shows that the estimate for all these factors is close to zero (the maximum being 0.23). Our findings therefore suggest that none of these parameters fully explain infants' preference for speech sounds.

In other words, from birth on, infants show a preference for speech, which cannot be reduced to the three simpler explanations tested here: naturalness, vocalness, or familiarity (represented here by comparing effects for native versus competitor against those for foreign versus competitor). It is possible that infants prefer speech because of its complex acoustic structure and fast transitions (Rosen & Iverson, 2007). Spectral or temporal modulations taken separately are not sufficient to elicit neural responses similar to the ones elicited by speech (Minagawa-Kawai, Cristià, Vendelin, Cabrol, & Dupoux, 2011). However, speech is characterized by joint spectrotemporal modulations at specific rates (Singh & Theunissen, 2003). It is possible that infants are sensitive to this specific spectro-temporal structure (though see Norman-Haignere & McDermott, 2018 showing that they only explain neural responses in primary auditory cortex, suggesting that other factors contribute to the behavioral response in later processing stages). Testing this explanation would require to compute the modulation spectra of the actual stimuli used in the experiments. Thus, we

recommend interested researchers to deposit their stimuli in a public archive such as the Open Science Framework (Foster & Deardorff, 2017).

A capacity to preferentially listen to speech sounds from birth suggests that infants are born with the capacity to recognize their conspecifics' communication signals. This parallels what has been proposed for faces: Infants are born with the capacity to orient their attention toward them, even without any prior exposure to faces (???; Morton & Johnson, 1991). This would stem from basic perceptual abilities present at birth, namely that the visual system would be tuned to a spatial structure that correspond to those of faces (???; Morton & Johnson, 1991). As newborns have never been exposed to such visual stimuli before, it would reflect general properties (i.e. filters) of the visual system. Similarly, the auditory system could be tuned to a spectro-temporal structure that speech presents. The combination of this non-specific bias with the systematic variations of the auditory environment would result in preferential responses to speech from birth. However, contrary to faces, fetuses are exposed to speech that is low-pass filtered by the womb throughout the last trimester of gestation (Lecanuet & Granier-Deferre, 1993; Querleu, Renard, Versyp, Paris-Delrue, & Crèpin, 1988). It is therefore possible that prenatal experience with low-pass filtered speech helps infants to form a representation of speech, by refining the response properties of the auditory system to speech. The fact that familiarity with the language used in the experiment did not modulate infants' preference suggests that this effect is not triggered by familiarity with the sounds of the native language. Infants would therefore form a representation that is specific enough to discriminate speech from other natural or vocal sounds, but general enough to be independent of the language spoken.

Ultimately, preferential processing of speech may support higher level cognitive tasks. The human species is a gregarious one. Detecting speech signals would allow to integrate it with other sensory percepts, such as faces, to form multisensory representations of conspecifics (Vouloumanos et al., 2009). This would lay the track for social cognition.

Identifying speech signals and paying attention to them would allow infants to form complex representations of the sensory world, that they can manipulate cognitively. Consistently, infants could categorize visual stimuli (i.e., associate a label to a category of objects) when they were associated to speech, but not pure tones or backward speech (Ferry, Hespos, & Waxman, 2010, 2013; Fulkerson & Waxman, 2007). Interestingly, infants categorized visual stimuli when presented with speech, melodies, monkey (Fulkerson & Haaf, 2003), or lemur vocalizations (Ferry et al., 2013). These results support the idea that infants may preferentially process complex sounds. Finally, the preference itself may also be a meaningful index of processing that can be used to identify children at risk (Sorcinelli et al., 2019). Understanding this phenomenon is therefore crucial for both theoretical and clinical advances.

TEXT TO INTEGRATE Newborns' brain activation was different for forward than backward speech when the native language was used as the speech stimuli, but not when a foreign language was used (May, Gervain, Carreiras, & Werker, 2018). neuro-imaging results failed to report differential response to speech, human non-communicative vocalizations, and rhesus calls in 1- to 4-month-old infants (Shultz, Vouloumanos, Bennett, & Pelphrey, 2014). These results are in line with adult neuroimaging data showing distinct responses to speech as compared to other natural or environmental sounds, even in low-level auditory regions (Norman-Haignere, Kanwisher, & McDermott, 2015; Norman-Haignere & McDermott, 2018).

We want to close by drawing the reader's attention to methodological insights the present study provides. First, the median sample size at present is 16, which is close to the field standard (Bergmann et al., 2018), but much lower than current recommendations (Oakes, 2017). Well-chosen sample sizes are crucial for powerful experiments (Simmons, Nelson, & Simonsohn, 2011). Our meta-analysis provides the average effect size across the literature, which will, in turn, allow researchers to run power analyses to determine the sample size they need in future experiments. Second, the distribution of effect sizes in the

literature is consistent with publication bias, in view of a strong asymmetry of the funnel plot. The trim-and-fill method suggested 11 points may be missing. Given that we have 54 effect sizes in total, a fifth more would be missing. The missing experiments are in the negative section, i.e., a preference against speech, a result that could lead authors to doubt their own data and not submit it to journals, or that would be considered odd by reviewers and editors, who may ask that the data be removed (or who may recommend the paper to be rejected altogether). The literature being biased toward positive effect sizes, the true effect size might be smaller than the one we found (vertical line on Figure 6). To correct this bias, we invite researchers to use registered reports (Kiyonaga & Scimeca, 2019). In this new publication scheme (available for Developmental Science, Infancy, Infant Behavior and Development, and Journal of Child Language at the time of writing, see a full up-to-date list on https://cos.io/rr/), manuscripts are submitted before data are collected. Reviewers and editors make publication decisions based solely on the introduction and methods. The paper is then reviewed once more for readability, but it cannot be rejected if the results are surprising or uncomfortable for the field. This would facilitate the publication of experiments failing to report a speech preference, or actually reporting a preference for the competitor, which would in turn help to draw a more accurate picture of the phenomenon. Our dataset can be community-augmented, and we invite researchers investigating this phenomenon to complement it with any data they would have (https://osf.io/4stz9/?view only=d0696591ebf34bfc8430f848cd945ca8), whatever the results

(https://osf.io/4stz9/?view_only=d0696591ebf34bfc8430f848cd945ca8), whatever the results and publication status.

Our meta-analysis revealed uneven distributions of experiments across age and stimulus dimensions. One may think that this uneven age distribution could be one potential reason for the lack of an age effect. However, by aggregating the numerous individual studies of a literature, meta-analyses gain statistical power. As such, meta-analytic results have more cumulative explanatory value than any single study. Previous experiments suggested that infants initially perceive speech and other vocal sounds as a common category, and then

narrow it to speech around five months of age. With a sample size of 1071 infants, covering a much wider age range than individual experiments, our meta-analysis provide evidence for an alternative view of the phenomenon, namely that infant prefer speech from birth onwards, even to other natural or vocal sounds, and that this effect is not driven by familiarity with the sounds of the language used. This clearly points to the value of meta-analysis: to advance a field and inspire follow-up studies. In particular, future experiments should test infants older than 9 months. Language production gains in complexity at about this age (Oller, Eilers, Neal, & Schwartz, 1999), which could affect infants' speech preference. Experiments using natural vocal stimuli as competitor, and foreign speech as target, would contribute to fill in the gap in the litterature that our meta-analysis revealed.

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 $\label{thm:continuous} \begin{tabular}{ll} Table 1 \\ Statistical \ results \ of \ meta-regression \ without \ any \ moderator. \end{tabular}$

	estimate	SE	t	confidence interval
average effect size	0.34	0.06	5.41	0.21 - 0.47

Table 2

Statistical results of meta-regression with all moderators. The intercept corresponds to the effect size when speech is in a foreign language, and the competitor is natural, and vocal, at age 0. The moderator estimates correspond to changes in the intercept when the target stimuli are in the native language (familiarity); the competitor is artificial (naturalness); and the competitor is non-vocal (vocal quality).

	estimate	SE	t	confidence interval
intercept	0.32	0.14	2.20	0.01 - 0.63
familiarity	-0.06	0.12	-0.48	-0.32 - 0.2
naturalness	0.23	0.11	2.02	-0.02 - 0.47
vocal quality	-0.15	0.15	-0.98	-0.48 - 0.18
age	0.00	0.00	0.29	0 - 0

Table 3

Statistical results of meta-regression with naturalness and its interaction with age as moderators. The intercept corresponds to the effect size when the competitor is natural, at age 0. The moderator estimates correspond to changes in the intercept when the competitor is artificial (naturalness).

	estimate	SE	t	confidence interval
intercept	0.32	0.21	1.52	-0.21 - 0.86
naturalness	0.04	0.23	0.15	-0.47 - 0.54
naturalness*age	0.00	0.00	0.39	0 - 0

Table 4

Statistical results of meta-regression with vocal quality and its interaction with age as moderators. The intercept corresponds to the effect size when the competitor is vocal, at age 0. The moderator estimates correspond to changes in the intercept when the competitor is non-vocal (vocal quality).

	estimate	SE	t	confidence interval
intercept	0.45	0.15	2.97	0.1 - 0.81
vocal quality	-0.33	0.19	-1.77	-0.74 - 0.07
vocal quality*age	0.00	0.00	1.47	0 - 0.01

Table 5

Statistical results of meta-regression with language of the speech sounds and interaction with age. The intercept corresponds to the effect size when speech is in a foreign language at age 0. The moderator estimates correspond to changes in the intercept when the target stimuli are in the native language (familiarity).

	estimate	SE	t	confidence interval
intercept	0.65	0.30	2.14	-0.06 - 1.35
familiarity	-0.39	0.32	-1.22	-1.09 - 0.32
familiarity * age	0.00	0.00	1.54	0 - 0.01

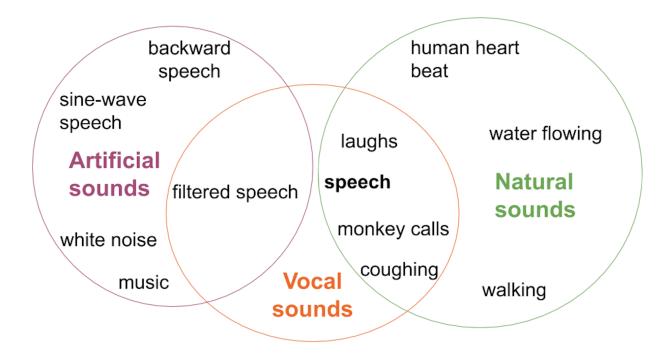


Figure 1. Speech is a natural, vocal sound.

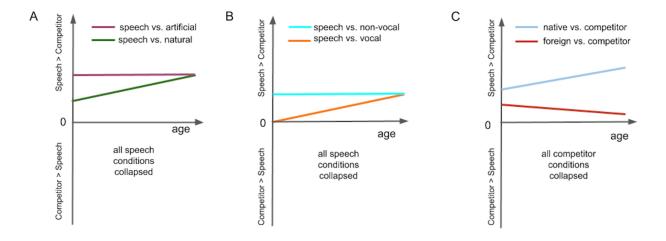


Figure 2. Hypothesized pattern of preference: the x axis shows age, the y axis represents the effect size derived from the contrast between a speech condition and a competitor condition (preference for speech over the competitor is plotted up; the lower quadrants are empty because we do not predict a preference for the competitor over speech). A: Speech contrasted to natural (green) or artificial (purple) competitors. B: Speech contrasted to vocal (orange) or non-vocal (cyan) competitors. C: Collapsing across competitors, separating speech in a foreign language (red); speech in the native language (blue).

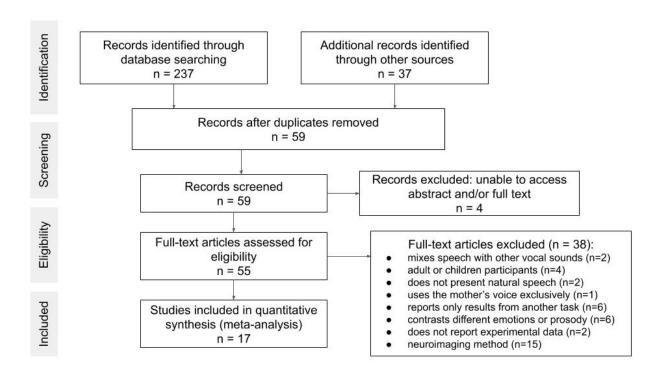


Figure 3. PRISMA flowchart summarizing the literature review and selection process.

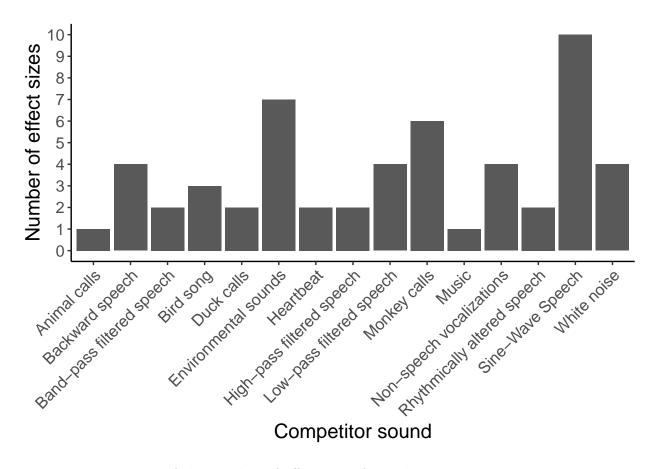


Figure 4. Histogram of the number of effect sizes for each competitor.

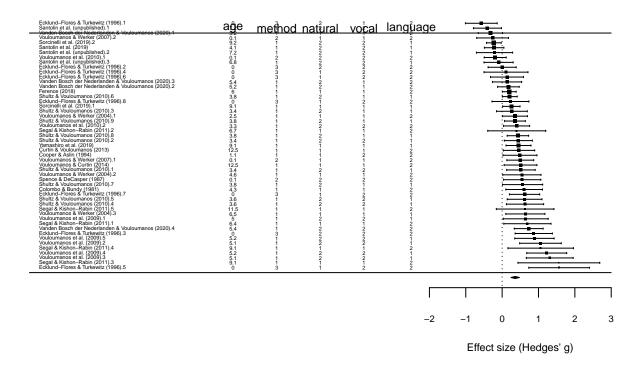


Figure 5. Forest plot of effect sizes available in the literature, along with their respective moderator status.

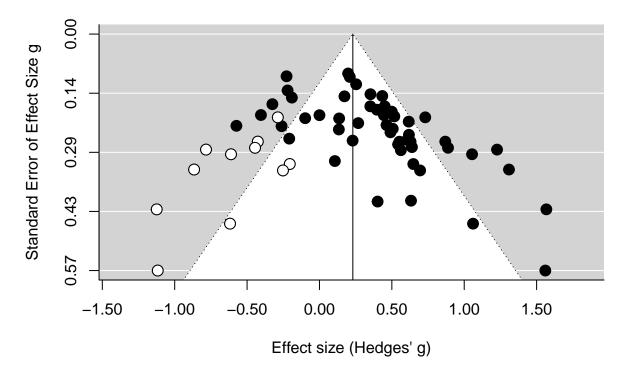


Figure 6. Funnel plot of effect sizes and their respective standard errors. Black dots: effect sizes observed in the literature. White dots: missing effect sizes, suggestive of a publication bias². Vertical line: average effect size after filling the missing effect sizes.

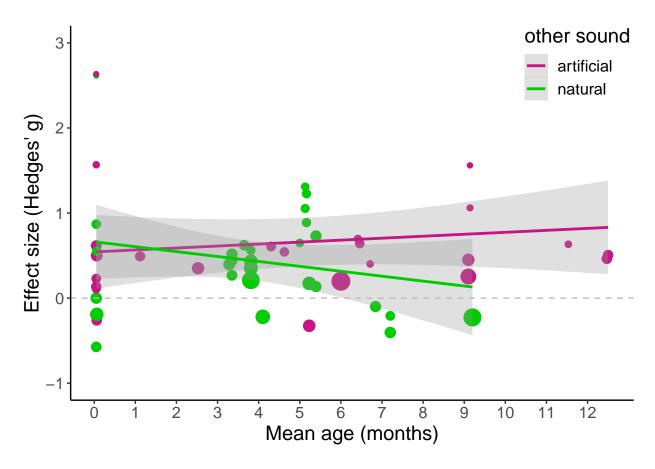


Figure 7. Effect sizes as a function of age and natural quality of the competitor. The size of each dot is inversely proportional to the variance. Positive effect sizes reflect a preference for the speech sound, negative effect sizes reflect a preference for the competitor sound.

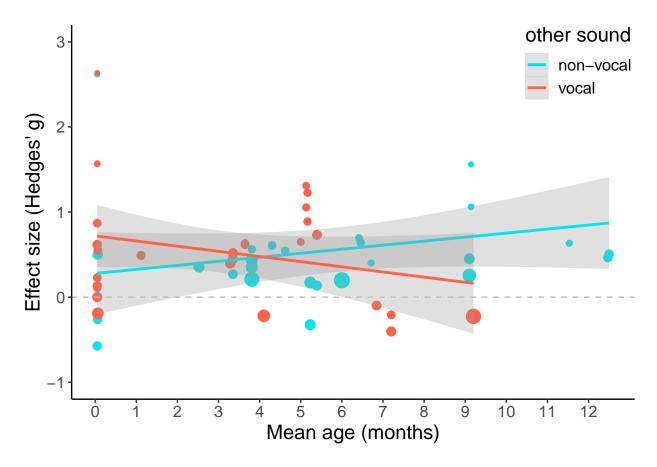


Figure 8. Effect sizes as a function of age and vocal quality of the competitor. The size of each dot is inversely proportional to the variance. Positive effect sizes reflect a preference for the speech sound, negative effect sizes reflect a preference for the competitor sound.

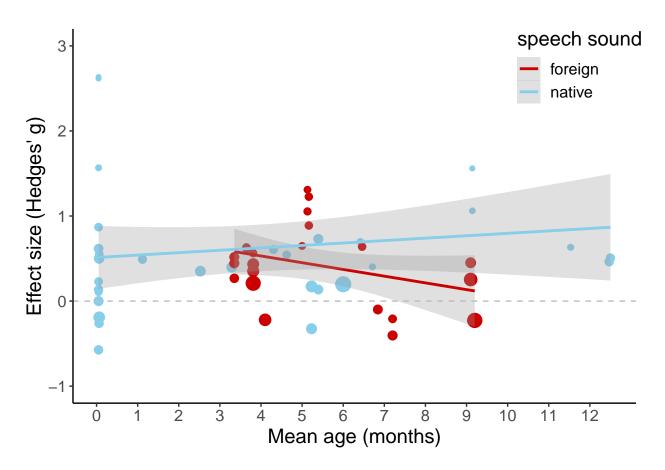


Figure 9. Effect sizes as a function of age and familiarity with the speech sounds. The size of each dot is inversely proportional to the variance. Positive effect sizes reflect a preference for the speech sound, negative effect sizes reflect a preference for the competitor sound.