Integration of visual information in auditory cortex promotes auditory scene analysis through

multisensory binding

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Abstract

How and where in the brain audio-visual signals are bound to create multimodal objects remains unknown. One hypothesis is that temporal coherence between dynamic multisensory signals provides a mechanism for binding stimulus features across sensory modalities in early sensory cortex. Here we report that temporal coherence between auditory and visual streams enhances spiking representations in auditory cortex. We demonstrate that when a visual stimulus is temporally coherent with one sound in a mixture, the neural representation of that sound is enhanced. Supporting the hypothesis that these changes represent a neural correlate of multisensory binding, the enhanced neural representation extends to stimulus features other than those that bind auditory and visual streams. These data provide evidence that early cross-sensory binding provides a bottom-up mechanism for the formation of cross-sensory objects and that one role for multisensory binding in auditory cortex is to support auditory scene analysis.

When listening to a sound of interest, we frequently look at the source. However, how auditory and

visual information are integrated into a coherent perceptual object is unknown. The temporal

properties of a visual stimulus can be exploited to detect correspondence between auditory and visual

streams [1-3], can bias the perceptual organisation of a sound scene [4], and can enhance or impair

listening performance depending on whether the visual stimulus is temporally coherent with a target

or distractor sound stream [5]. Together, these behavioural results suggest that temporal coherence

between auditory and visual stimuli can promote binding of cross-modal features to enable the

formation of an auditory-visual (AV) object [6].

Visual stimuli can both drive and modulate neural activity in primary and non-primary auditory cortex

[7-11] but the contribution that visual activity in auditory cortex makes to auditory perception remains

unknown. One hypothesis is that the integration of cross-sensory information into early sensory cortex

provides a bottom-up substrate for the binding of multisensory stimulus features into a single

perceptual object [6]. We have recently argued that binding is a distinct form of multisensory integration

that underpins perceptual object formation, and can be separated from other sorts of integration by

demonstrating a benefit in the behavioural or neural discrimination of a stimulus feature orthogonal

to the features that link crossmodal stimuli (Fig. 1a). Therefore, in order to demonstrate binding an

appropriate crossmodal stimulus should elicit not only enhanced neural encoding of the stimulus

features that bind auditory and visual streams, but that there should be enhancement in the

representation of other stimulus features associated with the source (Fig. 1c).

Here we test the hypothesis that the incorporation of visual information into auditory cortex can

determine the neuronal representation of an auditory scene through multisensory binding (Fig.1). We

demonstrate that when visual luminance changes coherently with the amplitude modulations of one

sound stream in a mixture, the neural representation of that sound stream is enhanced in the auditory

cortex. Consistent with these effects reflecting cross-modal binding, the encoding of auditory timbre,

an orthogonal stimulus dimension, is subsequently enhanced in the temporally coherent auditory

stream.

Results

We recorded neuronal responses in the auditory cortex of awake (n=9 ferrets, 221 single units, 311

multi-units) and medetomidine-ketamine anesthetised ferrets (n=5 ferrets, 426 single units, 772 multi

units) in response to naturalistic time-varying auditory and visual stimuli adapted from Maddox et al[5].

Recordings in anesthetized animals allowed us to isolate bottom-up attention-independent

processing, permitted longer recording durations for additional control stimuli and enabled

simultaneous characterization of neural activity across cortical laminae. Recordings in awake animals

while they held their head at a drinking spout but were not engaged in a behavioral task allowed us

to measure neural activity free from any confounds associated with pharmacological manipulation

and in the absence of task-directed attention. The auditory streams were two vowels with a distinct

pitch and timbre (denoted A1: [u], fundamental frequency (F0) = 175 Hz and A2: [a], F0 = 195 Hz) each

of which was independently amplitude modulated with a low-pass (<7 Hz) envelope (Fig. 1d). A full-

field visual stimulus accompanied the auditory stimuli, the luminance of which was temporally

modulated with a modulation envelope from one of the two auditory streams (Fig. 1e). We tested

stimulus conditions in which a single AV stimulus pair was presented ('single stream' stimuli), where

the auditory and visual streams could be temporally coherent (A1V1, A2V2) or independently

modulated (A1V2, A2V1). We also tested a dual auditory stream condition in which both auditory

streams were presented and the visual stimulus was temporally coherent with one of the auditory

streams (A12V1 or A12V2, Fig. 1e).

Spike patterns in auditory cortex differentiate dynamic auditory-visual stimuli

Before exploring the impact of temporal coherence between auditory and visual stimuli on auditory

cortical neurons, we first used the responses to single stream stimuli to classify neurons according to

whether they were modulated primarily by auditory or visual temporal dynamics. To determine

whether the auditory amplitude envelope reliably modulated spiking we used a spike-pattern classifier

to decode the auditory stream identity, collapsed across visual stimuli (i.e. we decoded auditory

stream identity from the combined responses to A1V1 and A1V2 stimuli and the combination of A2V1

and A2V2 responses). An identical approach was taken to determine if neuronal responses reliably

distinguished visual modulation (i.e. we decoded visual identity from the combined responses to A1V1

and A2V1 stimuli and the combined responses elicited by A1V2 and A2V2). Neuronal responses which

were informative about auditory or visual stimulus identity at a level better than chance (estimated

with a bootstrap resampling) were classified as auditory-discriminating (Fig. 2a-b) and / or visual-

discriminating (Fig. 2c-d) respectively.

In awake animals, 39.5% (210/532) of units were auditory-discriminating, 11.1% (59/532) were visual-

discriminating, and only 0.38 (2/532) discriminated both auditory and visual stimuli. Overall a smaller

proportion of units represented the identity of auditory or visual streams in the anesthetised dataset:

20.2% (242/1198) were auditory-discriminating, 6.8% (82/1198) were visual-discriminating, and 0.58%

(7/1198) discriminated both.

During recordings made under anaesthesia, we also recorded responses to noise bursts and white-

light flashes (both 100 ms) presented separately and together to map AV responsiveness in auditory

cortex (Bizley et al., 2007). Specifically, responsiveness was defined using a two-way ANOVA (factors:

auditory stimulus [on/off] and visual stimulus [on/off]) on spike counts measured during stimulus

presentation. We defined units as being sound-driven (main effect of auditory stimulus, no effect of

visual stimulus or interaction), light-driven (main effect of visual stimulus, no effect of auditory

stimulus or interaction) or both (main effect of both auditory and visual stimuli or significant

interaction; $p < 0.05$). Using such stimuli revealed that the classification of units by visual / auditory

discrimination of single stream stimuli selected a subset of light or sound driven units and that the

proportions of auditory, visual and AV units recorded in our sample were in line with previous studies

(Bizley et al, 2007: 65.1% (328/504) of units were driven by noise bursts, 16.1% (81/504) by light

flashes and 14.1% (71/504) by both.

We hypothesised that temporal coherence between auditory and visual stimuli would enhance the

discriminability of neural responses, irrespective of a unit's classification as auditory or visual

discriminating. We confirmed this prediction by comparing discrimination of temporally coherent

(A1V1 vs. A2V2) and temporally independent (A1V2 vs. A2V1) stimuli (Fig. 2e, f): Temporally coherent

AV stimuli produced more discriminable spike patterns than those elicited by temporally independent

ones in both awake (Fig. 2e, pairwise t-test, auditory-discriminating t_{418} = 34.277, p<0.001; visual-

discriminating t_{116} = 13.327, p<0.001; All t_{540} = 35.196, p<0.001) and anesthetised recordings (Fig.2f,

auditory-discriminating t_{482} = 27.631, p<0.001; visual-discriminating t_{162} = 22.907, p<0.001; All t_{664} =

33.149, p<0.001).

What might underlie the enhanced discriminability observed for temporally coherent cross-modal

stimuli? The phase of low frequency oscillations determines the excitability of the surrounding cortical

tissue[12-14], is reliably modulated by naturalistic stimulation[15-19] and has been implicated in

multisensory processing[20,21]. We hypothesised that sub-threshold visual inputs could modulate spiking

activity by modifying the phase of the local field potential such that phase coupling to temporally

coherent sounds is enhanced. This in turn would provide a mechanism by which neuronal spiking was

enhanced when auditory and visual streams are temporally coherent.

Dynamic visual stimuli elicit reliable changes in LFP phase

Stimulus evoked changes in the local field potential (LFP) were evident from the recorded voltage

traces and analysis of cross-trial phase coherence demonstrated that there were reliable changes in

phase across repetitions of identical AV stimuli (Fig. 3 a, b). To isolate the influence of visual activity

on the LFP for each unit, and address the hypothesis that visual stimuli elicited reliable changes in the

LFP, we calculated phase and power dissimilarity functions for stimuli with identical auditory signals

but differing visual stimuli[17]. Briefly, this analysis assumes that if the phase (or power) within a

particular frequency band differs systematically between responses to two different stimuli, then

inter-trial phase coherence (ITPC) across repetitions of a single stimulus will be greater than across

randomly selected stimuli. For each frequency band in the LFP, we therefore compared "within-

stimulus" ITPC for responses to each stimulus (A1 stream Fig. 3c; A2 stream Fig. 3d) with "across-

stimulus" ITPC calculated from stimuli with identical auditory stimuli but randomly selected visual

stimuli (e.g. randomly drawn from A1V1 and A1V2; Fig. 3c). The difference between within-stimulus

and across-stimulus ITPC was then calculated across frequency and described as the phase

dissimilarity index (PDI) (Fig. 3e, f) with positive PDI values indicating reliable changes in phase

coherence elicited by the visual component of the stimulus.

We calculated PDI values for each of the four single stream stimuli and grouped conditions by

coherency (coherent: A1V1 / A2V2, or independent: A1V2 / A2V1). To determine at what frequencies

the across-trial phase reliability was significantly positive, we compared within-stimulus and across-

stimulus PDI for each frequency band (paired t-test with Bonferroni correction for 43 frequencies, α =

0.0012). In awake subjects we identified a restricted range of frequencies between 10 and 20 Hz where

visual stimuli enhanced the phase reliability (Fig. 4a, b). In anesthetised animals, average PDI values

were larger than in awake animals and all frequencies tested had single stream PDI values that were

significantly positive (Fig. 4d, e). We therefore conclude that visual stimulation elicited reliable

changes in the LFP phase in auditory cortex. In contrast to LFP phase, a parallel analysis of across trial

power reliability showed no significant effect of visual stimuli on LFP power in any frequency band

(Supplementary Fig. 1).

Next we asked whether there were any frequencies at which phase coherence was increased by AV

temporal coherence by performing a pairwise comparison of single stream PDI values, yielded from

temporally coherent and independent stimuli, for all frequency points. In anaesthetised animals, the

single stream PDI did not differ between coherent and independent stimuli at any frequency (Fig. 4f).

In awake animals, PDI values were similar for temporally coherent and temporally independent

stimuli, except in the 11-14 Hz band where coherent stimuli elicited significantly greater phase

coherence (Fig. 4c). Together these data suggests that visual inputs modulate the phase of the low

frequency field potential in auditory cortex independently of temporal coherence with auditory

stimuli, and are consistent with auditory cortical neurons integrating visual and auditory information

such that discriminability of spiking responses to temporally coherent auditory visual signals are

enhanced (Fig.2e, f).

Visual information enhances the representation of the temporally coherent auditory stream in a

sound mixture

Arguably the greatest challenge for the auditory brain is to reconstruct sound sources in the world

from their overlapping cochlear representations. Having demonstrated that temporal coherence

between auditory and visual stimuli enhances discriminability of auditory spiking responses, we next

asked whether the temporal dynamics of a visual stimulus could enhance the representation of one

sound in a mixture. We therefore recorded responses to auditory scenes composed of two sounds

(A1 and A2) presented simultaneously with a visual stimulus that was temporally coherent with one

or other auditory stream (A12V1 or A12V2). To test if a visual stimulus could enhance the

representation of the temporally coherent auditory stream in such dual stream stimuli we then

compared dual stream responses with responses to temporally coherent single stream stimuli.

Figure 5 illustrates this approach for a single unit: responses to the single stream AV stimuli (Fig. 5a)

formed templates against which we judged the similarity of responses to the dual stream stimuli (Fig.

5b). Responses to the dual stream stimuli more closely resembled A1V1 when the visual stimulus was

V1, and A2V2 when the visual stimulus was V2. In our analysis of single-stream encoding, this unit was

classified as visual-discriminating, but many auditory-discriminating units showed similar response

properties (e.g. Supplementary Fig. 2). Enhancement of the coherent auditory stimulus representation

was visible at the population level (Fig. 5c-f): Auditory cortical responses to dual-stream stimuli most

closely resembled responses to the single stream stimulus with the same visual component. This

finding was robust in both awake (Fig. 5d, pairwise t-test: t_{540} = 6.073, p<0.001) and anesthetised

animals (Fig. 5f, t_{660} = 9.5137, p<0.001) suggesting that these effects were not mediated by attention.

Modulation of dual stream responses by visual stimulus identity was not simply a consequence of the

shared visual component of single stream and dual stream stimuli (Fig. 6). To show this we decoded

responses to dual stream stimuli (A12V1 and A12V2) using responses to auditory-only stimuli (Fig. 6a;

A1 or A2). We also analysed responses to mixed auditory streams with no visual stimulus (A12) using

responses either to coherent single stream stimuli (A1V1, A2V2). A two-way repeated measures

ANOVA on the decoder responses with factors of visual stream (V1, V2, no visual), and template type

(AV or A) demonstrated a significant effect of visual stream identity on dual stream decoding (Fig. 6d,

$F_{(2, 528)}$ = 19.320, p <0.001), but there was no effect of template type ($F_{(1, 528)}$ = 0.073, p = 0.787)

or interaction between factors ($F_{(2, 528)}$ = 0.599, p = 0.550). Post-hoc comparison across units

revealed that without visual stimulation there was no tendency to respond preferentially to either

stream but that visual stream identity significantly influenced classification of dual stream responses.

The ability of a visual stimulus to modulate auditory representation in the dual stream condition was

observed across all cortical layers (defined by current source density analysis, see online methods,

Supplemental Fig. 3a,b) and across three tonotopic fields (Supplemental Fig. 3c,d) of anaesthetized

subjects. While present in all effects, the influence of the visual stimulus was strongest in the supra-

granular layers (supplemental Fig.3b). Thus cross-modal modulation by temporal coherence was a

general phenomenon across auditory cortex. Separating auditory, visual, and auditory-visual units

according to responsiveness to classic neurophysiological stimuli (noise bursts and light flashes)

revealed that the impact of a visual stimulus on the representation of a sound mixture was present

across functional sub-populations but strongest in visual and auditory-visual units (Supplemental Fig.

4). Finally, we observed these effects in both single and multi-units (Supplemental Fig.5).

Visual stimuli elicit changes in LFP phase in the context of an auditory scene

Our findings indicate that visual stimuli can shape the representation of auditory mixtures and that

temporal coherence between auditory and visual stimuli enhances across-trial phase coherence. To

understand whether changes in phase coherence could provide a mechanism for visual modulation of

auditory representations, we again generated within-stimulus ITPC for each dual-stream stimulus (Fig.

7a, A12V1 and A12V2) and across-stimulus ITPC by randomly selecting responses across visual

conditions (Fig. 7b). We then expressed the difference as the dual stream PDI (dual stream phase

dissimilarity index, Fig. 7c). Since the auditory components were identical in each dual stream

stimulus, the influence of the visual component on LFP phase could be isolated as non-zero dual

stream PDI values (paired t-test, bonferoni corrected, $\alpha = 0.0012$). In awake animals, dual stream PDI

was significantly greater than zero at 11-14Hz and 16-19 Hz (Fig. 7d, e) whereas in anesthetised

animals, we found positive dual stream PDI values across all frequencies tested (Fig. 7f, g). In

anesthetised animals where we were able to use the responses of units to noise and light flashes to

categorise units as auditory, visual or auditory-visual, we also confirmed significant PDI values in each

of these subpopulations of units (Supplemental Fig. 4c). In awake animals, we confirmed the

significance of PDI values in the 11-14 Hz range across different rates of amplitude modulation of the

auditory stimulus (Supplemental Fig. 6).

Neural responses to auditory timbre deviants are enhanced when changes in visual luminance and

auditory intensity are temporally coherent

A hall-mark of an object-based rather than feature-based representation is that all stimulus features

are bound into a unitary perceptual construct, including those features which do not directly mediate

binding[22]. We predicted that binding across modalities would be promoted via synchronous changes

in auditory intensity and visual luminance (Fig. 1b) and observed that the temporal dynamics of the

visual stimulus enhanced the representation of temporally coherent auditory streams (Fig. 2e-f and

5d-f). To determine whether temporal synchrony of visual and auditory stimulus components also

enhanced the representation of orthogonal stimulus features (Fig. 1c) and thus fulfil a key prediction

of binding, we introduced brief timbre perturbations into our dual stream stimuli (n = 4 deviants, two

in A1 and two in A2). Such deviants could be detected by human listeners and were better detected

when the auditory stream in which they were embedded was temporally coherent with an

accompanying visual stimulus[5]. We hypothesised that, despite containing no information about the

occurrence of deviants, a temporally coherent visual stimulus would enhance the representation of

changes in timbre in the responses of auditory cortical neurons.

To isolate neural responses to the timbre change from those elicited by the on-going amplitude

modulation, we extracted the 200ms epochs of the neuronal response during which the timbre

deviant occurred and compared these to epochs from responses to otherwise identical stimuli without

deviants. We observed that the spiking activity of many units differed between deviant and no-deviant

trials (e.g. Fig 8a) and so we used a pattern-classifier approach to estimate the presence/absence of a

timbre deviant in a given response window. We first considered the influence of temporal coherence

between auditory and visual stimuli on the representation of timbre deviants in the single stream

condition (A1V1, A1V2 etc.). We found that a greater proportion of units detected at least one deviant

when the auditory stream in which deviants occurred was temporally coherent with the visual

stimulus relative to the temporally independent condition. This was true both for awake (Fig. 8b;

Pearson chi-square statistic, χ^2 = 322.617, p < 0.001) and anesthetised animals (Fig. 8e; χ^2 = 288.731, p

< 0.001). For units that discriminated at least one deviant, discrimination scores were significantly

higher when accompanied by a temporally coherent visual stimulus (Fig.8c, awake dataset, pairwise

t-test t_{300} = 3.599 p<0.001; Fig. 8f, anesthetised data t_{262} = 4.444 p<0.001).

Across the population of units, we performed a two-way repeated measures ANOVA on discrimination

performance with visual condition (V1/V2) and the auditory stream in which the deviants occurred

(A1/A2) as factors. We predicted that enhancement of the representation of timbre deviants in the

temporally coherent auditory stream would be revealed as a significant interaction term. Significant

interactions were seen in both the awake (Fig. 8d, $F_{(1, 600)}$ = 29.138, p<0.001) and anesthetised

datasets (Fig. 8g, $F_{(1, 524)} = 16.652$, $p<0.001$). We also observed significant main effects of auditory

and visual conditions in awake (main effect of auditory stream, $F_{(1, 600)} = 4.565$, $p = 0.033$; main effect

of visual condition, $F_{(1, 600)} = 2.650$, $p = 0.010$) but not anesthetised animals (main effect of auditory

stream, $F_{(1,524)} = 0.004$, $p = 0.948$; main effect of visual condition, $F_{(1, 524)} =1.355$, $p = 0.245$). Thus

we concluded a temporally coherent visual stimulus can enhance the representation of features (here

auditory timbre) orthogonal to those that promote binding between auditory and visual streams. This

finding is consistent with our model of cross-modal binding (Fig. 1a, c) and so these data fulfil our

definition of binding.

Discussion

Here we provide mechanistic insight into how auditory and visual information could be bound

together to form coherent perceptual objects. Visual stimuli elicit reliable changes in the phase of the

local field potential in auditory cortex that result in an enhanced spiking representation of auditory

information. These results are consistent with the binding of cross-modal information to form a

multisensory object. When two sounds are presented together within an auditory scene, the

representation of the stream that is temporally coherent with the visual stimulus is enhanced.

Importantly, this enhancement is not restricted to the encoding of the amplitude changes that bind

auditory and visual information but extends to the encoding of auditory timbre, a stimulus dimension

orthogonal to the dimensions that link auditory and visual stimuli. Thus our results meet the

requirements for a strict neural test of cross-modal binding that was laid out in Bizley et al.[6]. These

data provide a physiological underpinning for the pattern of performance observed in human listeners

performing an auditory selective attention task in which detection of a pertubation in a stimulus

stream is enhanced or suppressed, when a visual stimulus is temporally coherent with the target or

masker auditory stream respectively[5]. The electrophysiological data presented here suggest that the

temporally coherent auditory stream would be represented more effectively, making the task easier

when this stream was the target and making the task more challenging when it was the masker.

Surprisingly, the effects of the visual stimulus on the representation of an auditory scene can be

observed in anesthetised animals ruling out any top-down effect of attentional modulation.

Previous investigations of the impact of visual stimuli on auditory scene analysis have frequently used

speech stimuli. In order to probe more general principles that might relate to both speech and non-

speech processing we chose to employ non-speech stimuli, but utilized modulation rates that fell

within the range of syllable rates in human speech [23]. Previous work has demonstrated that a visual

stimulus can enhance the neural representation of the speech envelope both in quiet and in noise [3,

24, 25]. Being able to see a talker's mouth provides listeners with rhythm information and information

about the amplitude of the speech waveform which may help listeners by cueing them to pay

attention to the auditory envelope[26] as well as information about the place of articulation that can

disambiguate different consonants[27]. Visual speech information is hypothesised to be relayed in

parallel to influence the processing of auditory speech: Our data support the idea that early

integration of visual information occurs[26,28-30] and is likely to reflect a general phenomenon whereby

visual stimuli can cause phase-entrainment in the local field potential. Our data support the

contention that such early integration is unlikely to be specific to speech. Indeed low-frequency

entrainment to modulations in an on-going stimulus are observed in the human brain and have been

shown to optimize listening performance for non-speech stimuli [31]. In contrast, later integration is

likely to underlie information about speech gestures that might be used to constrain lexical identity[26].

Consistent with previous studies, our analysis of local field potential activity revealed that visual

information reliably modulated the phase of oscillatory activity in auditory cortex independently of

the modulation frequency of the stimulus[8-11]. Neuronal excitability varies with LFP phase[32-35] and may

be the physiological mechanism through which cross-sensory information is integrated. Our analysis

allowed us to isolate changes in LFP phase that were directly attributable to the visual stimulus and

identified reliable changes in the LFP phase irrespective of whether the visual stimulus was temporally

coherent with the auditory stimulus. Such a finding is consistent with the idea that the LFP phase

synchronization arises from fluctuating inputs to cortical networks[14,21,36]. Our finding that visual

stimulation elicited reliable phase modulation in both awake and anesthetised animals suggests that

bottom-up cross-modal integration interacts with selective attention, which also modulates phase

information in auditory cortex[20]. While our data suggest that cross-modal binding can occur in the

absence of attention, it is likely that the effects we observe in auditory cortex are the substrates on

which selective attention acts to further boost the representation of cross-modal objects.

In the awake animal the impact of visual stimulation on LFP phase reliability was smaller than in the

anesthetised animal and was restricted to a narrower range of frequencies, consistent with a

dependence of oscillatory activity on behavioural state[37-39]. Since the neural correlates of multisensory

binding are evident in the anesthetised animal, the specific increase in alpha phase reliability that

occurred in awake animals in response to temporally coherent auditory-visual stimulus pairs (Fig. 4c

& 7e) may indicate an attention-related signal triggered by temporal coherence between auditory and

visual signals. Phase resetting or synchronisation of alpha phase has been associated both with

enhanced functional connectivity[38] and as a top-down predictive signal for upcoming visual

information[40]. Disambiguating these possibilities would require simultaneous recordings in auditory

and visual cortex and/or recording during the performance of a task designed to explicitly manipulate

attention.

Temporal coherence between sound elements has been proposed as a fundamental organising

principle for auditory cortex[41, 42] and here we extend this principle to the formation of cross-modal

constructs. Our data provide evidence that one role for the early integration of visual information into

auditory cortex is to resolve competition between multiple sound sources within an auditory scene.

While previous studies have demonstrated a role for visual information in conveying lip movement

information to auditory cortex [3, 8, 9, 20], here we suggest a more general phenomenon whereby visual

temporal cues facilitate auditory scene analysis through the formation of cross-sensory objects. The

origin of the visual inputs is an open question but both visual cortical and sub-cortical structures

innervate tonotopic auditory cortex[7, 43]. Identifying which of these inputs is responsible for the

physiological effects we observe requires experiments that manipulate defined neural circuits.

In summary, activity in auditory cortex was reliably affected by visual stimulation in a manner that

enhanced the representation of temporally coherent auditory information. Enhancement of auditory

information was observed for sounds presented alone or in a mixture and for sound features that

were related to (amplitude) and orthogonal to (timbre) variation in visual input. Such processes

provide mechanistic support for a coherence based model of cross-modal binding in object formation.

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Author contributions

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Figure 1: Hypothesis and experimental design

a Conceptual model illustrating how binding can be identified as a distinct form of multisensory

integration. Multisensory binding is defined as a subset of multisensory integration which results in

the formation of a crossmodal object. During binding, all features of the audio-visual object are linked

and enhanced including both those features that bind the stimuli across modalities (here temporal

coherence between auditory (A) intensity and visual (V) luminance) and orthogonal features such as

auditory pitch and timbre and visual color and size. Other forms of multisensory integration would

result in only the features that promote binding being enhanced - here auditory intensity and visual

luminance. To identify binding therefore requires a demonstration that non-binding features (e.g.

here pitch, timbre, color or size) are enhanced **b** When two competing sounds (red and blue

waveforms) are presented they can be separated on the basis of their features, but may elicit overlapping neuronal representations in auditory cortex. **c** Hypothesised enhancement in auditory stream segregation when a temporally coherent visual stimulus enables multisensory binding. When the visual stimulus changes coherently with the red sound (A1, top) this sound is enhanced and the two sources are better segregated. Perceptually this would result in enhanced auditory scene analysis and an enhancement of the non-binding features. **d** Stimuli design: Auditory stimuli were two artificial vowels (denoted A1 and A2), each with distinct pitch and timbre and independently amplitude modulated with a noisy low pass envelope. **e** Visual stimulus: a luminance modulated white light was presented with one of two temporal envelopes derived from the amplitude modulations of A1 and A2. **f** illustrates the stimulus combinations that were tested experimentally in *single stream* (a single auditory visual pair) and *dual stream* (two sounds and one visual stimulus) conditions.

Figure 2: Auditory-visual temporal coherence enhances neural coding in auditory cortex.

A pattern classifier was used to determine whether neuronal responses were informative about auditory or visual stimuli. The responses to single stream stimuli are shown for two example units, with responses grouped according to the identity of the auditory (**a, b**, auditory discriminating unit)

or visual stream (**c, d**, visual discriminating unit). In each case the stimulus amplitude (a,b) / luminance

(c,d) waveform is shown in the top panel with the resulting raster plots (20 trials per condition) and

peri-stimulus time histogram (20 ms bin) below. **e, f**: Decoder performance (mean ± SEM) for

discriminating stimulus identity (coherent: A1V1 vs. A2V2; independent: A1V2 vs. A2V1) in auditory

and visual classified units recorded in awake (e) and anaesthetised (f) ferrets. Pairwise comparisons

for decoding of coherent versus independent stimuli: p<0.001 (***, see results).

Figure 3: Visual stimuli elicit reliable changes in the phase of the local field potential

a, b Example LFP responses to single stream stimuli across visual conditions. Data obtained from the

recording site at which multiunit spiking activity discriminated auditory stream identity in Fig. 2 a and

b. The amplitude waveforms of the stimuli are shown in the top panel, with the evoked LFP

underneath (mean across 21 trials). The resulting inter-trial phase coherence (ITPC) values are shown

in the bottom two panels. **c, d** ITPC was calculated for coherent and independent AV stimuli separately

and compared to a null distribution (ITPC across). Single stream phase dissimilarity values (PDI) were

calculated by comparing ITPC values to the ITPC across condition (**e, f**).

Figure 4: Visual stimuli elicit reliable changes in LFP phase in awake and anesthetised animals.

Mean inter-trial phase coherence (ITPC) values across frequency for coherent (**a, d**) and independent

(**b, e**) conditions. Dots indicate frequencies at which the ITPC values were significantly greater than

chance (Pairwise ttest, α = 0.0012, Bonferroni corrected for 43 frequencies). **c f:** Mean (±SEM) single

stream phase dissimilarity index (PDI) values for coherent and independent stimuli in awake (**c**) and

anaesthetised (f) animals. Black dots indicate frequencies at which the coherent stream PDI is

significantly greater than in the independent conditions (p<0.001).

Figure 5: Visual stimuli can determine which sound stream auditory cortical neurons follow in a mixture.

Spiking responses from an example unit (visual classified unit from the awake dataset) in response to **a,** single stream AV stimuli used as decoding templates and **b**, dual stream stimuli, rasters and PSTH responses stimuli from which single trial responses were classified. When the visual component of the dual stream was V1, the majority of trials were classified as A1V1 (82% (19/23) of trials), and A2V2 when the visual stimulus was V2 (26% (6/23) of responses classified as A1V1 (see also green datapoint in c).). **c-f** data for awake (c,d) and anesthetised (e,f) datasets. In each case the left panel (c,e) shows the distribution of decoding values according to the visual condition and the right panel (d,f) shows

 the population mean (± SEM) Pairwise comparisons revealed significant effect of visual conditionI on

 decoding in both datasets (p<0.001).

Figure 6: Visual stimuli shape the neural representation of an auditory scene.

In an additional control experiment (n=89 units recorded in awake animals), the responses to coherent

AV and auditory-only (A Only) single stream stimuli were used as templates to decode dual stream

stimuli either accompanied by visual stimuli (V1/V2) or in the absence of visual stimulation (no visual).

Spiking responses from an example unit in response to **a,** single stream stimuli (A only, no visual) used

as decoding templates and **b**, dual stream stimuli, in each case the auditory waveform, rasters and

PSTHs are shown. A two-way ANOVA with template type and visual condition as factors revealed

significant effects of visual condition but not template type. Post-hoc comparisons demonstrated that

classification was significantly influenced by visual stimulus identity when both A only and AV

templates were used. The proportion of responses classified as A1 when the visual stimulus was V1 or

V2 are shown in **c** and **d,** Mean (± SEM) values for these units. Pairwise comparisons revealed

significant effect across visual conditions in both datasets (p<0.001).

Figure 7: Visual stimuli elicits reliable changes in LFP phase to shape auditory scene analysis

a, Auditory stimulus waveform (top), evoked LFP (middle) and resulting inter-trial phase coherency

plots (bottom) for a typical recording site in response to dual stream stimuli. **b**, ITPC values were

calculated across frequency for responses to dual stream stimuli with identical visual stimuli ('ITPC

within') and across randomly drawn visual stimuli ('ITPC-across', grey). Dual stream phase selectivity

index (PDI) values were then calculated as the difference between shuffled and dual stream ITPC (**c).**

d, f Average ITPC (± SEM) for dual stream ITPC-within and ITPC-across for awake and anaesthetised

animals. Symbols indicate where the dual stream PDI was significant (pairwise t-test for ITPC within

versus ITPC across, α = 0.0012 with correction). **e, g** Mean (± SEM) dual stream PDI values for awake

and anaesthetised animals.

Figure 8: Temporally coherent changes in visual luminance and auditory intensity enhance the coding of a non-binding auditory feature.

a Example unit (from the awake dataset) showing the influence of visual temporal coherence on spiking responses to dual stream stimuli with or without deviants embedded. Shaded rectangles

indicate the 200 ms window over which the timbre deviant occurred and over which analysis was

conducted. **b-d** timbre deviant discrimination in the awake dataset. Two deviants were included in

each auditory stream giving a possible maximum of 4per unit **b,** Histogram showing the number of

deviants (out of 4) that could be discriminated from spiking responses   **c,** Box plots showing the

average timbre deviant discrimination scores in the single stream condition across different visual

conditions (Coh: coherent, ind: independent). The boxes show the upper and lower quartile values,

and the horizontal lines at their "waist" indicate the median. **d,** Discrimination scores for timbre

deviant detection in dual stream stimuli. Discrimination scores are plotted according to the auditory

stream in which the deviant occurred and the visual stream that accompanied the sound mixture. **e-g**

show the same as **b-d** but for the anesthetised dataset.

458 Supplementary Figures

Supplementary figure 1: Dynamic visual stimuli do not elicit reliable changes in LFP power.

a, b: Mean (±SEM) single stream phase dissimilarity index values for coherent (A1V1 and A2V2) and independent (A1V2 and A2V1) stimuli at all recording sites at which there was a driven spiking response for awake (a) and anaesthetised (b) animals.

Supplementary figure 2: Visual stimuli can determine which sound stream auditory cortical neurons follow in a mixture in auditory-discriminating units

The spiking responses of an example unit are shown to coherent single stream (**a**) and dual stream

stimuli (**b**). This example unit was an auditory discriminating unit recorded from an awake animal.

When the visual stimulus was V1, 67% of trials were classified as A1V1, when the visual stimulus was

V2 only 29% of trials were classified as A1V1.

Supplementary figure 3: Preference to the auditory stream with which the visual stimulus was

coherent observed across all cortical layers and in three tonotopic fields.

In the anesthetised animal, we determined tonotopic gradients from responses to pure tone stimuli

to accurately confirm the cortical field in which any given recording was made. Recordings were made

with linear shank electrodes, facilitating current source density analysis to identify the cortical layers.

a Proportion of responses classified as A1 when the visual stimulus was V1 or V2 are shown for

supragranular layers (SG, $t_{700}=5.686$, p<0.001), granular layer (G, $t_{878}=3.481$ p<0.001) and intragranular

layer (IG, $t_{690}=4.418$, p<0.001). Data shown as mean (± SEM) across units. A two-way ANOVA across

visual condition and cortical layers showed only a significant effect of visual condition (F (1, 2273)

=64.288, p<0.001) but not of layers (F (1, 2273) =0.91, p =0.404) and no interaction was observed (F

(1, 2273) =2.679, p=0.068). Significant post-hoc comparisons (p<0.05) are shown with an asterisks. **b,**

In order to compare the magnitude of the effect of the visual stimulus across cortical layers we

calculated visual modulation as the decoding score when the visual stimulus was V1 – the decoder

score when the visual stimulus was V2. Plotted are the mean (± SEM) visual modulation across

different layers. A one-way ANOVA across cortical layers showed a significant effect of cortical layer

($F_{(2, 1134)}$ =3.15, p =0.043). Post hoc comparison revealed that visual modulation in SG is greater

than G. **c,** as for a but for different cortical fields (A1, t_{798} =5.435, p<0.001; AAF, t_{636} =4.302, p<0.001;

PPF, t_{510} =3.609, p<0.001; PSF, t_{317} =0.932, p=0.352). A two way ANOVA across visual condition and

cortical field showed only a significant effect of visual condition (Visual condition: $F_{(1, 2267)}$ =40.301,

p<0.001; fields: $F_{(1, 2267)}$ =1.937, p=0.121; visual condition x fields: $F_{(1, 2267)}$ =1.804, p=0.144).

Significant post-hoc comparisons (p<0.05) are shown with an asterisks. In order to compare the

magnitude of the effects across cortical fields we again calculated a visual modulation value. **b,** Mean

(± SEM) visual modulation across different fields. A one-way ANOVA across cortical layers showed a

no significant effect of cortical field ($F_{(3, 1130)}$ =2.19, p =0.0877).

Supplementary figure 4: Auditory, visual and auditory-visual units are influenced by visual stimuli.

In anesthetised animals we also recorded responses to 100 ms noise bursts and/or LED flashes in order

to apply conventional classification of units as either auditory, visual or auditory-visual. Unit

classification was based on a two way ANOVA of spike counts calculated over a 200ms window

following stimulus onset with auditory and visual stimuli as factors. Units in which both auditory and

visual stimuli significantly modulated spiking or in which there was a significant interaction between

stimuli were classified as auditory-visual (AV). This analysis yielded recording sites at which units were

classified as auditory only (A; n=177 units), visual only (V, n=130) and AV (grey, n=150). This allowed

us to explore the visual modulation effects according to independently determined functional

definitions of units as auditory, visual or auditory visual.

a, The proportion of responses classified as A1 when visual stimuli as V1 or V2 is plotted with each

unit color coded according classification as A, V or AV. **b**, Mean (± SEM) values across the population.

Two way ANOVA revealed a significant effect of visual condition ($F_{(1, 1212)}$ = 173.463, p<0.001) and

unit type ($F_{(2, 1212)}$ = 1780.803 p<0.001), and interaction between factors ($F_{(2, 1212)}$ =88.66,

p<0.001) on the proportion of responses classified as A1. While the effect was strongest in AV and V

units, pairwise post-hoc comparisons revealed a significant effect of visual stimulus in all subgroups

(p<0.05). **c**, Mean (± SEM) dual stream phase dissimilarity index (PDI) values for recording sites

categorised according to the spiking responses recorded there. Symbols indicate where the dual

stream PDI index was significant (pairwise t-test, p<0.001 with correction). While the phase effects

are greatest at the sites where visual activity was recorded, significant dual stream PDI values were

observed in all three unit types. In all three cases significant phase coherence was seen at 12Hz,

13.5Hz-14.5Hz and 42.5-44.5Hz. Modulation at 10-12 Hz was only observed at sites in which AV and V

responses were recorded.

Supplementary figure 5: The effects of temporal coherence on single stream decoding and of

visual identity on dual stream decoding are evident in both single and multi-units.

a,b Discrimination accuracy for single stream decoding is indistinguishable for single units and multi-

unit activity. Two way ANOVA revealed a significant effect of visual condition and no effect of unit

type in discrimination accuracy values in awake recording (a, Visual condition: F(1,851) = 14.6607

p<0.01, Unit Type: F(1,851)= 1.02, p=0.312; interaction: F(1,851)=0.005, p=0.449) and anaesthetised

recording. (b, Visual condition: F(1,1847) = 24.514, p<0.01, Unit Type: F(1,1847)= 0, p=0.986;

interaction: F(1,1847)=0.66, p=0.418) **c, d**. Impact of visual stimulus on dual stream decoding in single

units and multi units. A two way ANOVA revealed only significant effect of visual condition and no

effect of unit type in discrimination accuracy values in awake recording (c, Visual condition: F(1,449)

= 19.41 p<0.01, Unit Type: F(1,851)= 1.07, p=0.414; interaction: F(1,851)=0.33, p=0.568) and

anaesthetised recording. (d, Visual condition: F(1,1847) = 15.41, p<0.01, Unit Type: F(1,1847)= 1.1,

p=0.506;interaction: F(1,1847)=0.66, p=0.418).

Supplementary figure 6: Dual stream PDI values in the alpha range are independent of amplitude modulation rate.

Stimuli were generated with three different amplitude modulation rates (<7Hz, as in the main experiment, <12Hz and <17Hz) and the responses to these were recorded in 92 units. Symbols indicate where the dual stream phase selectivity index was significant (pairwise t-test, p<0.05 with correction). In all three cases significant phase coherence is seen between 10Hz-11.5Hz, 19Hz-20Hz and 24-26 Hz.

Supplementary figure 7: Stimulus duration does not influence the frequency range across which significant single stream PDI values and dual stream PDI values are observed.

We observed that phase coherence values were higher for longer duration stimuli and therefore in

order to directly compare anesthetised and awake datasets, all analysis in the anesthetised dataset

was restricted to the first three seconds of stimuli. At longer stimulus durations the ITPC profile and

resulting PDI varied more smoothly with frequency (a, c); However at both durations phase values

were significantly different from zero at all frequencies. The pattern of significant phase selectivity

values was also preserved across stimulus durations. (**b, d**). Frequency points at which the single

stream PDI value and dual stream PDI values were similar in 3 second length (**a, b**) and 14 second

length (**c, d**) Blue, red and black symbols indicate where the PDI was significant (pairwise t-test, α =

0.0012 with correction).

Online Methods

Animal preparation: The experiments were approved by the Committee on Animal Care and Ethical Review of University College London and The Royal Veterinary College, and performed under license from the UK Home Office and in accordance with the Animals Scientific Procedures Act 1986. Neural responses were recorded in a total of 11 awake pigmented adult female ferrets (Mustela putorius furo; 1-5 years old). Data from 9 of these animals was used for the main experiment (532 units), data from 8 other animals (6/9 and two additional ferrets, 89 units) was collected for additional control analysis (Figures 6e, supplemental figure 6). Animals were chronically implanted with recording electrodes and passively listening/watching stimuli while holding their head at a water spout and receiving continuous water reward. These animals were trained in various listening tasks for other studies. An additional 5 adult females were used to record responses under anaesthesia.

Full methods for recording under anesthesia can be found in Bizley et al., [44]. Briefly, ferrets were anesthetized with medetomidine (Domitor; 0.022mg/kg/h; Pfizer, Sandwich, UK) and ketamine (Ketaset; 5mg/kg/h; Fort Dodge Animal Health, Southampton, UK). The animal was intubated and the left radial vein was cannulated in order to provide a continuous infusion (5 ml/h) of a mixture of medetomidine and ketamine in lactated ringers solution augmented with 5% glucose, atropine sulfate (0.06 mg/kg/h; C-Vet Veterinary Products) and dexamethasone (0.5 mg/kg/h, Dexadreson; Intervet, UK). The ferret was placed in a stereotaxic frame in order to implant a bar on the skull, enabling the subsequent removal of the stereotaxic frame. The left temporal muscle was largely removed, and the suprasylvian and pseudosylvian sulci were exposed by a craniotomy, revealing auditory cortex (Kelly et al., 1986). The dura was removed over auditory cortex and the brain protected with 3% agar solution. The eyes were protected with zero-refractive power contact lenses. The animal was then transferred to a small table in a sound-attenuating chamber. Body temperature, end-tidal CO_2, and the electrocardiogram were monitored throughout the experiment. Experiments typically lasted

between 36 and 56 h. Neural activity was recorded with multisite silicon electrodes (Neuronexus

Technologies) in a 1x 16, 2x 16 or 4x 8 (shank x number of sites) configuration.

Full surgical methods for recording implanting electrode arrays to facilitate recording from awake

animals are available in Bizley et al. [45]. Briefly, animals were bilaterally implanted with WARP-16 drives

(Neuralynx, Montana, USA) loaded with high impedance tungsten electrodes (FHC, Bowdoin, USA)

under general anaesthesia (medetomidine and ketamine induction, as above, isoflurane maintenance

1-3%). Craniotomies were made over left and right auditory cortex, a small number of screws were

inserted into the skull for anchoring and grounding the arrays, and the WARP-16 drive was anchored

with dental acrylic and protected with a capped well. Animals were allowed to recover for a week

before the electrodes were advanced into auditory cortex. Pre-operative, peri-operative and post-

operative analgesia were provided to animals under veterinary advice.

Stimulus presentation: All stimuli were created using TDT System 3 hardware (Tucker-Davis

Technologies, Alachua, FL) and controlled via MATLAB (Mathworks, USA). For recordings in awake

animals, sounds were presented over two loud speakers (Visaton FRS 8). Water deprived ferrets were

placed in a dimly lit testing box (69 x 42 x 52 cm length x width x height) and received water from a

central reward spout located between the two speakers. Sound levels were calibrated using a Brüel

and Kjær (Norcross, GA) sound level meter and free-field ½-inch microphone (4191). Auditory streams

were presented at 65 dB SPL (Fig. 1a). Visual stimuli were delivered by illuminating the spout with a

white LED which provided full field illumination (Precision Gold N76CC Luxmeter, 0 to 36.9 lux). The

animals were not required to do anything other than maintain their heads in position at the spout

where they were freely rewarded. Recording was terminated when animals were sated.

For anesthetised recordings, acoustic stimuli were presented using Panasonic headphones (Panasonic

RP-HV297, Bracknell, UK) at 65 dB SPL. Visual stimuli were presented with a white Light Emitting Diode

(LED) which was placed in a diffuser at a distance of roughly 10 cm from the contralateral eye so that

it illuminated virtually the whole contralateral visual field.

Stimuli and data acquisition: *Auditory stimuli* were artificial vowel sounds that were created in

Matlab (MathWorks, USA). In the behavioural experiment that motivated this study [5], stimuli were 14

seconds in duration. However, we adapted the stimulus duration in awake recordings to 3 seconds in

order to collect sufficient repetitions of all stimuli, and to ensure animals maintained their head

position facing forwards for the whole trial duration. In the anesthetised recording stimulus streams

were 14 seconds long, as in the human psychophysics [5] but we only analysed the first 3 seconds to

ensure datasets were directly comparable (Supplemental Figure 7).

Stimulus A1 was the vowel [u] (formant frequencies F1-4: 460, 1105, 2857, 4205 Hz, F0= 195Hz), A2

was [a] (F1-4: 936, 1551, 2975, 4263 Hz, F0= 175Hz). Streams were amplitude modulated with a noisy

lowpass (7 Hz cutoff) envelope. Unless specifically noted, the timbre of the auditory stream remained

fixed throughout the trial. However, we also recorded responses to auditory streams that included

brief timbre deviants. As in our previous behavioural study, deviants were 200ms epochs in which the

identity of the vowel was varied by smoothly changing the first and second formant frequencies to

and from those identifying another vowel. Stream A1 was morphed to/from [ε] (730, 2058, 2857,

4205 Hz) and A2 to/from [i] (437, 2761, 2975, 4263 Hz).

Visual stimuli were generated using an LED whose luminance was modulated with dynamics that

matched the amplitude modulation applied to A1 or A2. In single stream conditions a single auditory

and single visual stream were presented (e.g. A1V1, A1V2, A2V1, or A2V2) whereas in dual stream

conditions both auditory streams were presented simultaneously, accompanied by a single visual

stimulus (A12V1, A12V2, A12V1 A12V2) (Fig. 1e). Auditory streams were always presented from both

speakers so that spatial cues could not facilitate segregation, and stimulus order was varied pseudo-

randomly. In the anesthetised recordings each stimulus was presented 20 times. In the awake dataset,

where recording duration was determined by how long the ferret remained at the central location

(mean repetitions: 20, minimum: 14, maximum: 34).

During anaesthetised recordings, pure tone stimuli (150 Hz to 19 kHz in 1/3-octave steps, from 10 to

80 dB SPL in 10 dB, 100 ms in duration, 5 ms cosine ramped) were also presented. These allowed us

to characterize individual units and determine tonotopic gradients, so as to confirm the cortical field

in which any given recording was made. Additionally broadband noise bursts and diffuse light flashes

(100 ms duration, 70 dB SPL) were presented and used to classify a stimulus as auditory, visual or

auditory visual. LFPs were subjected to current source density analysis to identify sources and sinks as

described by Kaur et al. [46]

Data Analysis: Electrophysiological data were analysed offline. Spiking activity and local field potential

signals were extracted from the broadband voltage waveform by filtering at 0.3-5kHz and 1-150 Hz

respectively. Spikes were detected, extracted and then sorted with a spike-sorting algorithm

(WaveClus) (Quiroga et al., 2004).

We used a Euclidean distance based pattern classifier (Schnupp et al., 2006) with leave-one-out cross

validation to determine whether the neuronal responses to different stimuli could be discriminated.

Spiking responses to a given stimulus were binned into a series of spike counts from stimulus onset (0

s) to offset (3s) in 20 ms bins. The average across-repetition response to each stimulus (excluding the

to-be-classified response) were used as templates and the response to a single stimulus presentation

was classified by calculating the Euclidean distance between itself and the template sweeps and

assigning it to the closest template. To determine whether the classifier performed significantly better

than expected by chance, a 1000 iteration permutation test was performed where trials were drawn

(with replacement) from the observed data and randomly assigned to a stimulus that was then used

for template formation / decoding. A neural response was considered to be significantly informative

about stimulus identity if the observed value exceeded the 95th percentile of the randomly-drawn

distribution.

This approach allowed us to classify units according to their functional properties: auditory units

discriminated two auditory stimuli based on the amplitude modulation of sound (A1 versus A2)

regardless of visual dynamics, (Fig. 2a, b), visual units discriminated visual presentations based on

temporal envelope of visual stimuli (V1 versus V2) regardless of auditory presentation (Fig. 2c, d) and

AV units could do both. This approach was extended to classify dual stream responses by using the

average response to each of the temporally coherent single stream stimuli (A1V1 or A2V2) as

templates. Performance was (arbitrarily) expressed as the proportion of responses classified as being

from the A1, and compared for the two dual stream stimuli with different visual conditions (Figure 5).

To be considered in this analysis the response of a unit had to be informative about the single stream

stimuli (i.e. classified as either auditory and/or visual discriminating).

Phase/power dissimilarity analysis: Local field potential recordings were considered for all sites at

which there was a significant driven spiking response, irrespective of whether that response could

discriminate auditory or visual stream identity. For the single stream trials, we computed a single

Stream Phase Dissimilarity Index (PDI), which characterizes the consistency and uniqueness of the

temporal phase/power pattern of neural responses to continuous auditory stimuli (Luo and Poeppel,

2007). This analysis compares the phase (or power) consistency across repetitions of the same

stimulus with a baseline of phase-consistency across trials in which different stimuli were presented.

In the first stage of PDI analysis, we obtained a time-frequency representation of each response using

wavelet decomposition with complex 7-cycle Morlet wavelets in 0.5 steps between 2.5–45 Hz,

resulting in 86 frequency points. Next, we calculated the inter-trial phase-coherence value (ITPC;

Equ.1) at each time-frequency point, across all trials in which the same stimulus was presented. For

each frequency band, the ITPC time-course was averaged over the duration of the analysis window

and across all repetitions to obtain the average *within-stimulus ITPC*.

$$ITPC_{t,f} = \left| \frac{\sum_{k=1}^{N} e^{i\theta_{k,t,f}}}{N} \right|$$ Equ.1

In which N is equal to the number of trials, and θ is the phase of trial *k* at a given frequency (*f*) and

time (*t*). The *across-stimuli ITPC* was estimated using the same approach but using shuffled data, such

that the ITPC was computed across trials with the same auditory stimulus but randomly drawn visual

stimuli. The single stream phase dissimilarity index (Single stream PDI) was computed as the difference

between the ITPC value calculated for *within* trials and the ITPC values calculated *across* visual trials

(Equ.2). y. The dissimilarity function for each frequency bin *i* was defined as;

$$Single\ Stream\ PDI_i = \frac{\sum_{j=1}^{N} ITPC_{ij}\ within_{vis}}{N} - \frac{\sum_{j=1}^{N} ITPC_{ij}\ ,across_{vis}}{N}$$ Equ.2

Large positive PDI indicate that responses to individual stimuli have a highly consistent response on

single trials. Single stream PDI values were calculated for each stimulus type and then averaged across

stimuli to calculate values for temporally coherent and temporally independent auditory visual stimuli.

Single stream PDI was positive if within stimulus ITPC was larger than across-stimulus ITPC (pairwise

t-test, $p < 0.05$ Bonferroni correction for 86 frequencies points) and was considered significant if a

minimum of 2 adjacent bins exceeded the corrected threshold.

Dual stream phase dissimilarity index (dual stream PDI) values were calculated by extending this

approach for dual stream stimuli with the goal of determining how the temporal envelope of the visual

stimulus influences the neural response to a sound mixture. To this end, we calculated the *within-dual*

ITPC from the A12V1 trials and A12V2 trials separately and *across-dual ITPC* by randomly selecting

trials from both stimuli (Equ.3). The within-dual and across-dual ITPCs were then averaged over time

and subtracted to yield the dual stream PDI (Equ.3).

$$Dual\ Stream\ PDI_i = \frac{\sum_{j=1}^{N} ITPC_{ij}\ within_{dual}}{N} - \frac{\sum_{j=1}^{N} ITPC_{ij}\ across_{dual}}{N} \qquad \text{Equ.3}$$

Positive dual stream PDI values indicate that the time course of the neural responses was influenced

by visual input, despite the identical acoustic input. We determined whether the dual stream PDI was

greater if the *within_dual ITPC* was significantly larger than *across_dual ITPC* (pairwise t-test, $p < 0.05$

Bonferroni correction, as above).

Timbre deviant analysis*:* In order to determine how a visual stimulus influenced the ability to decode

timbre deviants embedded within the auditory streams we used the cross-validated pattern classifier

described above for analysing single stream stimuli to discriminate deviant from no-deviant trials.

Responses were considered over the 200 ms time window that the deviant occurred (or the equivalent

point in the no-deviant stimulus) binned with a 10 ms resolution. Significance was assessed by a 1000

iteration permutation test in which trials were randomly drawn with replacement from deviant and

no-deviant responses. The discrimination score was calculated as the proportion of correctly classified

trials.

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