Training enhances the ability of listeners to exploit visual information for

auditory scene analysis

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Abstract

The ability to use temporal relationships between cross-modal cues facilitates perception and

behavior. Previously we observed that temporally correlated changes in the size of a visual

stimulus and the intensity in an auditory stimulus influenced the ability of listeners to perform an

auditory selective attention task (Maddox et al., 2015). In this task participants detected timbral

changes in a target sound while ignoring those in a simultaneously presented masker. When the

visual stimulus was temporally coherent with the target sound, performance was significantly

better than when it was temporally coherent with the masker sound, despite the visual stimulus

conveying no task-relevant information. Here, we trained observers to detect audiovisual

temporal coherence and asked whether this improved their ability to benefit from visual cues

during the auditory selective attention task. We observed these listeners improved performance

in the auditory selective attention task and changed the way in which they benefited from a

visual stimulus: after training performance was better when the visual stimulus was temporally

coherent with either the target or the masker stream, relative to the condition in which the visual

stimulus was coherent with neither auditory stream. A second group which trained to

discriminate modulation rate differences between temporally coherent audiovisual streams

improved task performance, but did not change the way in which they used visual information. A

control group did not change their performance between pretest and post-test. These results

provide insights into how crossmodal experience may optimize multisensory integration.

Keywords: audiovisual integration, selective attention, auditory scene analysis, temporal

processing, training

1 Introduction

Building an accurate perception of the world around us requires that the brain appropriately link

signals across modalities (Kayser & Shams, 2015). Auditory and visual signals arrive and are

processed with different latencies. Consequently, cross modal signals can be perceived as

synchronous across a range of onset asynchronies – known as the temporal binding window

(Dixon & Spitz, 1980; Meredith, Nemitz, & Stein, 1987). Previous studies have demonstrated

that there is short term plasticity in this window (Megevand, Molholm, Nayak, & Foxe, 2013;

Navarra et al., 2005; Schormans & Allman, 2018; Vroomen, Keetels, de Gelder, & Bertelson,

2004; Zmigrod & Zmigrod, 2015) and that experience and longer-term training can narrow this

window such that listeners more accurately judge synchronous from asynchronous stimuli

(Bidelman, 2016; Dixon & Spitz, 1980; H. Lee & Noppeney, 2011; Powers, Hillock, & Wallace,

2009).

Training listeners to refine their temporal binding window has been observed to have varied

consequences for multisensory integration. For example, training which was effective in

narrowing the temporal window for spatial ventriloquism also decreased the likelihood of cross-

modal interactions as indexed by spatial ventriloquism (McGovern, Roudaia, Newell, & Roach,

2016). In contrast, training improved visual temporal discrimination abilities, but did not

influence the likelihood of perceiving sound induced flash illusions (Ryan A. Stevenson, Wilson,

Powers, & Wallace, 2013). Finally, in another study in which listeners were trained to

discriminate asynchronous from synchronous stimuli listeners were subsequently shown to have

stronger spatial ventriloquism effects when auditory-visual signals were temporally synchronous

but spatially separated (Sürig, Bottari, & Röder, 2018).

A problem in interpreting these varied effects is that many lab-based tasks do not well capture

the complexity that the brain faces in real-world situations. In most lab-based paradigms

observers often judge single audio and visual signals, presented in an otherwise quiet and dark

environment. In contrast, in the world, brain must match one of several competing sounds to a

given visual object (or vice versa). Moreover, due to the variance in the timing of real-world

signals, simply narrowing the window over which integration occurs may be a suboptimal

strategy for effective information integration. Rather, what observers need to do is detect

whether temporal coherence exists between signals in different modalities so that they may be

appropriately grouped (Lee, Maddox, & Bizley, 2019).

An additional consideration for training studies that focus on the temporal binding window is

that it is unclear to what extent any adjustments in the temporal binding window extend to other

multisensory processing tasks. In many cases the task used to train observers is the same one

used to measure the temporal binding window raising the question of how generalizable results

are and whether they represent a genuine change in cross-modal binding, or whether listeners are

simply shifting an internal criterion in order to improve their performance in this one task

(Bizley, Maddox, & Lee, 2016).

In this study our goal was to ask whether we could train listeners to improve their ability to

detect audiovisual temporal coherence and, in doing so, whether this would improve their ability

to use visual cues to appropriately group sound elements from one stream and separate them

from those elements in a competing sound. To train listeners to detect audiovisual temporal

coherence listeners were required to differentiate streams in which audio and visual stimuli were

amplitude / radius modulated in a statistically independent manner from stimuli in which audio

and visual elements maintained some degree of temporal coherence. We elected to train listeners

to detect small amounts of correspondence (rather than detect incoherence) as we reasoned that

detecting moments of genuine correspondence is more likely to be useful than detecting

incoherence.

In addition to measuring the ability of observers to assess temporal coherence before and after

training (using similar stimuli to those used in the training), we used the auditory selective

attention task from Maddox et al., (2015) to assess how effectively listeners could utilize visual

information during the performance of an auditory selective attention task. This task required

that subjects focused on one of two competing auditory streams and reported brief timbre

perturbations within the target stream. They also watched a visual stimulus whose radius could

change in a manner that was temporally coherent with either the target, the masker or neither

auditory stream (but was never predictive about the timing of the timbre perturbations). In

Maddox et al., we reported that the visual coherence condition significantly influenced

performance in the auditory selective attention task such that performance was better when the

visual stimulus was coherent with the target masker stream than when it was coherent with the

masker stream, In order to determine whether any training effects we observed were critically

dependent on improved temporal coherence detection, as opposed to passive exposure to

temporally coherent auditory-visual streams, we also trained another group of observers in an

amplitude modulation rate discrimination task with the same temporally coherent stimuli. A

control group simply performed the pretest and post-test without any training. We hypothesized

that an improved ability to detect temporal coherence may enable listeners to appropriately group

temporally coherent audiovisual streams which in turn would promote more effective auditory

selective attention. Our results support this hypothesis and demonstrate that only listeners trained

to detect audiovisual temporal coherence change the way in which they are able to use visual

information to augment auditory scene analysis.

Methods

Subjects

42 adults (age range 18–34 years; mean age 28 years; 11 males) with normal hearing and normal

or corrected-to-normal vision, participated in the study. Six participants were excluded after the

pretest due to poor performance (mean d'<0.8, n = 4), or low visual hit rates (<70%, n=2). The

remaining 36 participants were included for further analysis and were randomly allocated to 3

groups (12 listeners per group). The study was approved by the Ethics Committee of the

University College London (ref: 5139) and all procedures performed were in accordance with the

Declaration of Helsinki. All individuals were paid for their participation and signed an informed

consent form before participation.

Stimuli and testing procedure

Each participant performed a "pretest" and "post-test" that consisted of the auditory selective

attention task and an AV coherence detection threshold test. The pretest and post-test sessions

each lasted 90 minutes and were separated by a maximum of 2 weeks and a minimum of 5 days.

Participants in the control group did no training sessions but performed the pretest and post-test

within 2 weeks. Participants in the two training groups performed 5 training sessions on 5

separate days over not more than 2 weeks (Figure 1A).

Auditory selective attention task

The auditory selective attention task required that listeners attend to one of two competing

auditory streams and report the presence of brief (200 ms) timbre perturbations in the target

audio stream. They were additionally required to monitor a visual stimulus whose radius changed

in time. The two audio streams were independently amplitude modulated and the visual radius

was modulated with a time course that matched one or the other auditory stream or was

independent of them both (Figure 1B). Envelopes for visual envelope and auditory amplitude

were created using the same frequency domain synthesis. For each trial, an envelope was created

by first setting all amplitudes of frequency bins above 0 Hz and below 7 Hz to unity and others

to zero. At an audio sampling rate of 24,414 Hz, all non-zero bins were given a random phase

from a uniform distribution between 0 and 2π, the corresponding frequency bins across Nyquist

frequency were set to the complex conjugates to maintain Hermitian symmetry, and the inverse

Fourier transform was computed yielding a time domain envelope. Second and third envelopes

were created using the same method, and orthogonalized using a Gram-Schmidt procedure. Each

envelope was then normalized so that it spanned the interval [0, 1] and then sine-transformed [y

$= \sin2 (\pi x/2)$] so that the extremes were slightly accentuated. Visual envelopes were created by

subsampling the auditory envelope at the monitor frame-rate of 60 Hz, starting with the first

auditory sample, so that auditory amplitude corresponded with the disc radius at the beginning of

each frame.

Figure 1: A Experimental design. **B** Schematic representation of auditory and visual stimuli used in the auditory selective attention test (panel was taken from Maddox et al., 2015). Amplitude envelopes of target (black) auditory stream and masker (red) auditory stream and visual radius envelopes for three auditory visual (AV) coherence conditions; target coherent (black), masker coherent (red) and neither (blue). Examples frames of the visual stimuli at three radius level. **C** Schematic representation of AV temporal coherence detection test/ AVTC training. Two 5 seconds AV pairs were used. One maintained some degree of temporal coherence (left, here fully coherent) while the other was always fully independent (right) **D** Schematic representation of AV amplitude modulation rate discrimination (AVAM) training. Two 5 second temporally coherent AV pairs were used. Each pair had a different modulation envelope (and rate) but stimuli were always temporally coherent across modalities.

2   Stimuli were presented in an unlit sound-attenuating room over headphones (HD 555,

3   Sennheiser, Wedemark, Germany). Subjects were seated 60 cm from the screen with their heads

4   held stationary by a chinrest. Auditory stimuli were created in MATLAB and presented using an

RP2 signal processor (Tucker–Davis Technologies, Alachua, FL, USA). Each began and ended

with a 10 ms cosine ramp. All stimuli were presented diotically. Visual stimuli were synthesized

in MATLAB (The Mathworks, Natick, MA, USA) and presented using the Psychophysics

Toolbox (Brainard, 1997). The visual stimuli were gray discs that subtended between 1° and

2.5°. The white ring extended 0.125° beyond the gray disc.

The auditory stimuli were generated as described in the *timbre* variant of Maddox et al., (2015).

Each auditory stream was generated as a periodic impulse train and then filtered with synthetic

vowels simulated as four-pole filters (formants F1–F4). The /u/ stream had formant peaks F1–F4

at 460, 1105, 2857, 4205 Hz and moved slightly towards /ɛ/ during timbre events, with formant

peaks at 730, 2058, 2857, 4205 Hz. The /a/ stream had formant peaks F1–F4 at 936, 1551, 2975,

4263 Hz and moved slightly towards /i/ during timbre events, with formant peaks at 437, 2761,

2975, 4263 Hz. During timbre events the formants moved linearly toward the deviant for 100 ms

and then linearly back for 100 ms. Streams were calibrated to be 65 dB SPL (RMS normalized)

using an artificial ear (Brüel & Kjær, Nærum, Denmark) and presented against a low level of

background noise (54 dB SPL). Unlike Maddox et al., we did not assess individual timbre

discrimination thresholds but instead used a fixed level of difficulty determined using the

average of the individual thresholds measured previously. For [e] deviants in [u] stimuli this

corresponded to a shift of 42 Hz in F1 frequency and 143 Hz for F2, and for [i] deviants in [a]

stream there was a shift of 75 Hz for F1, 196 Hz for F2. Both vowels could take either F0 value

(175 Hz or 195 Hz, counterbalanced) and were equally likely to be target or masker.

Trials lasted 14 seconds. They began with only the target auditory stimulus and the visual

stimulus, indicating the to-be-attended (target) auditory stream to the subject. The to-be-ignored

auditory stream (masker) began 1 second later. As with the rest of the trial, the visual stimulus

was only coherent with the auditory target during the first second if it was a match-target trial.

All streams ended simultaneously. Events did not occur in the first two seconds (i.e. 1 second

after the masker began) or the last 1 second of each trial, or within 1.2 s of any other events in

either modality. A response made within 1 second following an event was attributed to that

event. To ensure audibility and equivalent target to masker ratios without providing confounding

information to the subject, an event in either auditory stream or the visual stream could only

begin when both auditory envelopes were above 70% maximum. There were between 1 and 3

inclusive events (mean events = 2) in both the target and masker in each trial. There were also

between 0 and 2 inclusive visual flashes per trial (mean flashes = 1), in which the outer ring

changed from white to cyan (0% red, 100% blue, 100% green) and back. Each subject completed

32 trials of each temporal coherence conditions (96 totals), leading to 64 potential hits and 64

potential false alarms for each condition (i.e., 128 responses considered for each d′ calculation)

as well as 32 visual flashes per condition. When computing d′, auditory hit and false alarm rates

were calculated by adding 0.5 to the numerator and 1 to the denominator so that d′ had finite

limits.

Auditory-visual temporal coherence detection test

The AV temporal coherence detection test was used to determine perceptual thresholds for

detecting AV temporal coherence. Auditory and visual stimuli were generated as described in the

auditory selective attention task above, with a single auditory and visual stream presented on

each occasion. The pitch and timbre of stimuli were varied across trials such that the auditory

stimuli were either [u] or [e] (without any timbre deviants embedded), with F0 = 175 or 195 Hz,

counterbalanced. The sounds in both stimulus intervals within the trial had identical pitch and

timbre values and had a duration of 5 seconds. In one stimulus interval, the sound was

accompanied by a visual stimulus in which the radius changed over time independently of the

auditory stimulus. In the other intervals, the auditory and visual stimulus maintained some

degree of temporal coherence (Figure 1C). The method of constant limits was used to determine

the threshold with subjects performing 20 trials at each coherence level. AV stimuli were

generated from 100% coherent in 10% steps to 10% coherent by multiplying the temporally

coherent envelope with an independent envelope. Participants were required to select the interval

(by pressing 1 or 2 on a button box) in which the temporally coherent pair was presented.

Feedback was provided on every trial.

Auditory-visual temporal coherence (AVTC) training

The stimuli and procedure in the AVTC training were identical to those used in the threshold

test, but with an adaptive three-down one-up rule to determine the coherence level of the

stimulus in the next trial. Previous work has demonstrated that task difficulty is an important

aspect in driving multisensory learning (De Niear, Koo, & Wallace, 2016) so by using this

approach we required that participants worked near to their threshold for a large proportion of

the training session. In the first training session, the stimuli in the first trial were 100% coherent,

and 100% independent. For the first 6 reversals coherency was decreased in 10% steps followed

by 5% steps for the following six reversals and by 2.5% steps for the remainder. The procedure

was terminated at 18 reversals unless a maximum of 150 trials was reached first. For the 2nd-5th

training session the first "coherent" stimulus was generated with the average coherence level of

the last ten reversals in the previous session. Each training session lasted less than 40 minutes.

Feedback was provided on every trial. Participants performed 5 training sessions on separate

days over not more than 14 days.

Amplitude modulation rate discrimination (AVAM) training

In AVAM training, two temporally coherent AV stimuli (5 seconds long, with a constant pitch

and timbre within a trial, counterbalanced across trials as described in "auditory-visual temporal

coherence detection test", Figure 1D) were sequentially presented. In one interval, the envelope

was always generated with a 7 Hz cut off rate, whereas the other was generated with a higher

rate (maximum AM cut off rate = 11 Hz). In the sessions of AM rate training, an adaptive three-

down one-up rule was used to determine the AM rate of the stimulus in the next trial. In the first

session, the first stimulus was generated at the maximum AM rate and differed in AM rate by

1Hz for the first six reversals and 0.5 Hz for the next six reversals and 0.25Hz for the rest of the

trials. The procedure was terminated at 18 reversals unless a maximum of 150 trials was reached

first. In each consequent session the first stimulus was generated with the average coherence

level of the last ten reversals in the previous session. Participants pressed "1" or "2" on the press

box to indicate the interval of the faster AV pair. Feedback was provided on each trial.

Statistical Analysis

Statistics were performed using MATLAB (2011b, Mathworks, USA) and SPSS (IBM). The d',

hit rates, false alarm, and visual hit rates across AV coherence conditions and pretest versus post-

test were calculated and statistical significance across groups was assessed by two-tailed

unpaired Student's t-tests, Mann-Whitney U-tests, one-way analysis of variance (ANOVA), mix

ANOVA or repeated measures of ANOVA where appropriate. Mauchly's test of sphericity was

used for the assumption of homogeneity of variance for independent tests in repeated measured

analysis if not otherwise reported. Significant main effects or interactions were followed up with

post-hoc testing using Bonferroni corrections where applicable. Significance was declared at p

<0.05, with a precise p value stated in each case, and all tests were two-sided. Partial eta

squared was reported for different ANOVA procedures in SPSS to demonstrate the effect size we

observed. For individual differences in AV temporal coherence detection thresholds, 95%

confidence intervals were calculated with a linear regression.

8   **Results**

We recruited participants and randomly assigned them to one of three groups, each of which

performed a pretest and a post-test. Pretests and post-tests comprised the timbre variant of the

selective attention task in Maddox et al., (2015) and an AV temporal coherence detection

threshold test. In between the pretest and post-test, one group trained on an AV temporal

coherence task (AVTC group, n = 12), one group trained on an AM rate discrimination task

using temporally coherent AV stimuli (AVAM group, n = 12), and a third group simply

performed the pretest and post-test separated by a minimum of 5 days (control, n = 12; Figure

1A).

We first confirmed that both trained groups improved their ability on the trained stimulus

feature. Figure 2A and B shows the thresholds derived from the last 5 reversals for the training

session on day 1 and day 5 for the AVTC group and the AVAM group respectively. For both

groups' thresholds were significantly lower for session 5, than for session 1 (pairwise t-test on S1

and S5 AVTC thresholds, $t_{11} = 2.961$, $p = 0.007$; AVAM thresholds: $t_{11} = 4.529$, $p<0.001$).

Next, we asked whether there was a change in performance in the auditory selective attention

task between the pretest and the post-test. We anticipated that participants improve their

performance between the pretest and the post-test in both trained groups, as these listeners had

been exposed to the vowel sounds during their training sessions (although the sounds that they

heard did not contain any timbre deviants).

Figure 2: Training improved performance in trained tasks and the Auditory Selective Attention task. **A** Training in auditory visual temporal coherence detection (AVTC) task was effective at driving an improvement in coherence detection threshold between session (S1) and session 5 (S5). Black vertical lines show the mean ± SEM across participants, gray lines are individual subjects. **B** Training in the auditory visual amplitude modulation (AVAM) task was effective at driving an improvement in AM rate discrimination between S1 and S5. **C** Both trained groups improved their overall d' in the auditory selective attention task. Mean d' across AV coherence condition of pretest and post-test for the three groups. *indicates significant differences (Pairwise t-tests p < 0.05)

To detect overall changes in performance that occurred independently of the AV coherence

condition, we calculated the overall sensitivity (d'), and directly compared the pretest and post-

test data across the three experimental groups (Figure 2C). We ran an 3x3x2 mixed ANOVA

for *d'*, with a between-subjects factor of experimental groups (AVTC, AVAM and control) and

within-subjects factors of sessions(pretest and post-test) and audio–visual coherence condition

(target coherent, masker coherent, neither, data available as source data 1).

We found a significant effect of session ($F_{(1, 33)} = 35.411$, $p < 0.001$, partial $\eta^2 = 0.518$), with

no significant effect of experimental group ($F_{(2, 33)} = 1.030$, $p = 0.368$, partial $\eta^2 = 0.059$) and

no significant interaction ($F_{(2, 33)} = 2.858$, $p = 0.072$, partial $\eta^2 = 0.148$). Post-hoc pairwise

comparisons between pretest and post-test data revealed that means of d' scores were

significantly different for both trained groups (AVTC group: $t_{11} = 3.065$, $p = 0.006$, AVAM

group: $t_{11} = 1.920$, $p = 0.034$; control group: $t_{11} = 0.854$, $p = 0.402$). Thus, both experimental

groups, but not the control group, improved their ability to detect timbre deviants in a sound

mixture.

We found a significant effect of coherence condition ($F_{(2, 66)} = 13.859$, $p < 0.001$, partial $\eta^2 =$

0.296) and significant interaction between session and coherence condition ($F_{(2, 66)} = 3.302$, p

$= 0.043$, partial $\eta^2 = 0.091$). In Maddox et al., (2015) we observed that performance in the

auditory selective attention task varied with AV coherence condition, with performance when the

visual condition was coherent with the target stream being superior to when the visual condition

was coherent with the masker stream. Focusing on the pretest data obtained from naïve listeners,

Bonferroni post-hoc tests showed that participants detect deviants significantly better in only

target coherent condition than masker coherent condition, thus replicating our original

observations.

Having observed that training increased performance on the auditory selective attention task, and

that prior to training listeners showed a characteristic pattern of performance across AV

coherence conditions, we went on to explore in more detail the effect of training on the way in

which the visual condition influenced performance in the auditory selective attention task. To do

this, we considered the d' values across condition, and, for visualization purposes, calculated the

normalized d' values (note: all statistical comparisons are performed on the untransformed

values). Normalised d' illustrates how performance varies across visual condition by taking the

difference between each condition and the across-condition mean. We hypothesized that if

training had no influence on the way in which listeners were influenced by the coherence

condition we would see the same across-coherence-condition pattern, simply shifted up to higher

d' values. In this situation the normalized d' values would be unchanged by training indicating

that the increase in d' was uniform across coherence condition. In contrast, if there was a change

in the magnitude of the visual stimulus induced effects we predicted a similar pattern of across-

coherence-condition performance, but a larger difference between target and masker coherent

conditions. This would be reflected in normalized d' measures being larger in magnitude, but

equivalent in sign between pretest and post-test. Finally, if training influenced the way in which

listeners utilized visual information, we predicted that there would be a change in the pattern of

across-coherence-condition d' values, and a change in the sign of the normalized values.

AVTC but not AVAM training alters the way listeners use visual cues in selective attention

task

We therefore calculated d', hit rates, false alarm rates and bias for each coherence condition in

the auditory selective attention task, for each training group and the pretest and post-test

separately. To analyse changes between pretest scores and post-test scores for each group, we

performed a two-way repeated measure within-subjects ANOVA with factors of coherence

condition (target coherent, masker coherent and neither) and session (pretest and post-test; Table

1). In this framework, we expect a significant effect of coherence condition in all groups and of

training in the AVTC and AVAM group. Of particular interest is the interaction term, as this

would indicate a training-induced change in the way in which visual information impacted

performance.

1   In t

Figure 3: Training to detect AV temporal coherence changed how listeners utilized visual information. **A-C** pretest (dashed line) and post-test (solid line) performance in the auditory selective attention task according to AV coherence conditions. A, D, G: AVTC group, B, E, H: AVAM group, C, F, G: control group. **D-F** Normalized mean ± SEM performance (calculated as within condition d' normalized to across condition performance for pretest and post-test separately). **G-I** the difference in d' between pre and post-test for each participant in either gray, or red. Participants who showed a greater increase in the masker coherent condition than the target coherent after training are plotted in red, the group mean is plotted in bold black.

1   temporal coherence of auditory and visual streams. Prior to training, participants showed the

1   expected effect of coherence condition (i.e. only target coherent > masker coherent). However,

after training a different pattern was observed: both target and masker coherent conditions were

superior to the neither condition in which the visual stimulus changed independently (Figure

3A). Notably, the normalized d' measures for the masker coherent condition changed from

negative to positive in this group (Figure 3D). Consistent with these observations a repeated

ANOVA analysis of d' scores revealed significant effect of session (F (1, 11) = 36.245, p

<0.001), a borderline effect of coherence condition (F (2, 22) = 6.908, p=0.05) and a significant

interaction (Figure 3A, D, G; F (2, 22) = 7.258, p = 0.002; see also Table 1 which reports effect

sizes). Post-hoc comparisons across AV coherence condition in the pretest data revealed that

participants performed better when the visual stimulus was coherent with the target auditory

stream versus the masker auditory stream (target coherent > masker coherent). In contrast, post-

Table 1: The results of two-way repeated measures within-subjects ANOVA for each variable (p < 0.05 in bold) for d', hit rates, false alarm and bias for three experimental groups.

	Coherence conditions			Session (pre vs post)			Interaction between coh. condition & session		
	F	p	η^2	F	p	η^2	F	p	η^2
AVTC Group									
d'	6.908	**.050**	.386	36.245	**<.001**	.767	7.258	**.002**	**.493**
Hit Rates	5.080	**.015**	.316	7.731	**.018**	.413	4.660	**.021**	**.298**
False Alarm*	2.692	.115	.197	17.164	**.002**	.609	2.555	.100	.189
Bias	1.233	.311	.101	0.128	.727	.110	3.005	.070	.215
AVAM Group									
d'	5.723	**.010**	.342	23.134	**.001**	.678	.009	.854	.014
Hit Rates	2.952	.073	.212	24.148	**<.001**	.687	.386	.682	.042
False Alarm*	2.317	.145	.174	5.846	**.034**	.347	.376	.655	.033
Bias*	1.909	.190	.148	3.337	.095	.233	.922	.411	.077
Control									
d'	4.653	**.021**	.297	1.431	.257	.115	.039	.962	.004
Hit Rates	0.998	.385	.083	1.180	.301	.097	.065	.929	.006
False Alarm	3.023	.069	.216	0.088	.772	.008	.253	.779	.022
Bias	0.301	.743	.027	0.084	.777	.008	.043	.958	.004

*Mauchly's test of sphericity was significant (<0.05), Greenhouse-Geisser corrected values are

reported.

hoc comparisons of the post-test d' scores revealed that, after training, performance was better

when the visual stimulus was coherent with either the target or the masker stream than in the

condition which coherent with neither (target coherent > neither, masker coherent > neither,

Bonferoni corrected post-hoc comparison p<0.05).

Participants in AVAM group were asked to detect the amplitude modulation rate of temporally

coherent AV pairs. They were not actively detecting temporal coherence, but passively exposed

to temporally coherent AV pairs. Although training improved their overall performance in the

auditory selective attention task, the way in which they used visual information appeared

unchanged after training. In both pretest and post-test performance was the best in the target

coherent condition (Figure 3B) and the normalized d' measures, which effectively factor out the

overall d' improvement, were overlapping (Figure 3E) suggesting that there was no change in the

way in which the coherence condition effected performance (Figure 3H). Both session ($F (1, 11)$

$= 23.134$, $p = 0.001$) and coherence condition ($F (2, 22) = 5.723$, $p = 0.010$) influenced d', but –

importantly – there was no interaction ($F (2, 22) = 0.09$, $p = 0.854$). Post-hoc comparisons

($p<0.05$) across coherence conditions revealed that subjects performed better when the visual

stimulus was coherent with the target auditory stream compared to the masker coherent condition

in both the pretest and post-test (target coherent > masker coherent). Therefore, this suggested an

overall improvement in performance after AVAM training, but no change in the way in which

subjects were able to exploit visual cues.

Performance in the control group did not differ significantly between pretest and post-test

(Figure 3C, F, I): there was no effect of session ($F (1,11) = 1.431$, $p = 0.257$), but a significant

Figure 4: Improved performance in the masker coherent condition in AVTC group was driven by a drop in false alarms and an increase in the masker-coherent condition hit rate. The changes in hit rates (**A**) and false alarm rate (**B**) between pretest and post-test; mean ± SEM.

effect of coherence condition ($F_{(2, 22)} = 4.653$, $p = 0.021$), with no interaction ($F_{(2, 22)} =$

0.039, $p = 0.962$). Participants performed significantly better when the visual stimulus was

coherent with the target auditory stream versus the masker auditory stream (target coherent >

masker coherent) in pretest and post-test (Figure 3F).

To better understand the effect in the AVTC group we considered the hit rates and false alarm

rates (which together define d') to determine whether the changes were principally driven by an

improved ability to detect the target timbre deviations in the masker coherent condition, or an

improved ability to ignore deviants that occurred in the masker stream. Figure 4 shows the

changes in hit rates and false alarm rates between the pretest and post-test for the AVTC group

and suggests that training drove an overall drop in false alarms across all coherence conditions,

and a condition-specific increase in hit rates, with the largest increase occurring in the masker

coherent condition. Two-way ANOVAs (table 1) revealed that there was a significant increase in

hit rates with training, and a significant session ($F_{(1, 11)} = 7.731$, $p = 0.018$), coherence

condition ($F_{(2, 22)} = 5.080$, $p = 0.015$), and interaction ($F_{(2, 22)} = 4.660$, $p = 0.021$). The

decrease in false alarm rate between pretest and post-test was also statistically significant (F

(1,11) = 17.164, p = 0.002) without a significant coherence condition effect (F (2, 22) = 2.692, p

= 0.115).

AVTC training improved temporal coherence detection

The pretest and post-test included a temporal coherence detection threshold test for all listeners.

As anticipated, those listeners trained to detect audiovisual temporal coherence improved their

thresholds between the pretest and post-test: (Figure 5A, t_{11} = 3.081, p = 0.005). Listeners in the

AVAM group who were exposed to temporally coherent AV stimuli did not improve their

thresholds (t_{11} = 1.69, p = 0.104) or not in the control group (t_{11} = 0.234, p = 0.817). Examination

of these data also revealed there was considerable variability in how well observers could detect

temporal coherence, and in the AVAM group, considerable individual variability in the way in

which performance changed between tests. We therefore asked whether any of this individual

variability predicted performance in the auditory selective attention task.

Figure 5: Training decreased temporal coherence thresholds in the AVTC group. AV temporal coherence threshold values of pretest and post-test for three groups.
* indicates significant paired t-test comparison (p<0.05).

15   **Sensitivity to temporal coherence is not predictive of performance in naïve listeners**

16   We first explored whether individual differences in temporal coherence detection ability

17   accounted for the impact that the coherence condition had on performance in naïve listeners.

Specifically, we tested the hypothesis that the ability of naïve listeners to detect AV temporal

coherence would predict their ability to benefit from AV temporal coherence in the auditory

selective attention task. We correlated each listener's AV temporal coherence threshold with the

difference between the d' score in the target and masker coherent condition (Figure 6A).

Contrary to this hypothesis, there was no relationship between these values (r = 0.1543, p =

0.3688), nor was there any relationship between overall performance (across condition d') and

AV temporal coherence thresholds (r = 0.2888, p = 0.0882).

Having observed that the AVTC group improved their ability to utilize audiovisual temporal

coherence in the masker coherent condition, we considered whether temporal coherence

thresholds might be correlated with the magnitude of the benefit / impairment that the masker

coherent condition had over the neither condition. To assess this, we considered the difference in

d' between the masker coherent condition and the neither condition. This comparison was

weakly negatively correlated with AV temporal coherence thresholds for naïve listeners (Figure

6B; r = 0.339, p = 0.0438) suggesting a trend where participants with better AV temporal

coherence thresholds were more able to exploit the temporal coherence between masker stream

and visual stimulus to yield a performance benefit relative to the neither condition. This finding

mirrors the effect of training whereby improving AV coherence thresholds led to an

improvement in the masker coherent condition.

Figure 6 Individual differences in AV temporal coherence detection
A Target coherent -masker coherent d' difference (n=36 naïve listeners) versus AV temporal coherence threshold (low values indicate better thresholds) from all listeners pretest data. Error bars indicate 95% confidence intervals **B** Masker-neither d' difference versus AV coherence threshold for the pretest data. **C** The change in overall performance between the pretest and post-test versus change in AV coherence threshold.

2   Finally, since some participants in all groups showed improved temporal coherence detection

3   thresholds, we asked whether at an individual subject level whether the change in temporal

4   coherence detection threshold correlated with the overall change in d': Participants with a larger

5   change in their AV coherence threshold showed larger improvements in overall performance

6   (Figure 4C; r = 0.353, p = 0.0347).

Discussion

Here we demonstrate that five short training sessions can improve a listener's ability to detect AV temporal correspondence and change the way in which they are able to exploit cross-modal temporal coherence. In naïve listeners a visual stimulus that is temporally coherent with a target auditory stream enhances performance relative to when the visual stimulus is temporally coherent with the masker stream (with a condition which was coherent with neither yielding intermediary performance). After training, both target and masker visual coherence conditions yielded significantly better performance relative to the condition in which the visual stimulus was coherent with neither.

We had two control groups in this study. The first did not perform any training in between the pretest and the post-test. The second group was trained on a temporal rate discrimination task that utilized temporally coherent auditory visual stimuli. Therefore, like the AVTC group they judged temporal features of the stimuli and were exposed to the stimulus streams that formed the target sounds in the auditory selective attention task. Unlike the AVTC group, the AVAM group did not require that observers make a cross-modal discrimination and observers were free to base their decisions on auditory and/or visual features. Consistent with perceptual learning resulting from exposure to the sounds, both groups improved in their performance of the auditory selective attention task. However, only observers that were required to explicitly judge cross-modal temporal coherence showed a change in the way in which visual information was used for auditory scene analysis.

We have previously argued (Maddox et al., 2015) that the visual stimulus impact performance in

the auditory selective attention task by altering how well listeners are able to separate the two

competing streams and select the target. This is because the features that link the audio and

visual streams (temporally coherent changes in auditory intensity and visual size) are

independent of the changes in sound timbre that listeners are required to detect and the visual

stimulus itself conveys no information about whether or when (or in which stream) the auditory

timbre deviants occurred. Thus, improved performance in the auditory task demonstrated that

auditory and visual streams have been bound into a single perceptual object (Bizley et al., 2016;

Lee et al., 2019). Recordings in the auditory cortex of passively listening, naïve ferrets

demonstrate that audiovisual temporal coherence causes an enhanced representation of the

temporally coherent stream that extends to all of its features (Atilgan et al., 2018). As well as

providing further evidence for audiovisual object formation, assuming that an audiovisual object

has processing advantages, or captures selective attention more strongly, these data provide a

bottom-up explanation for how the enhanced performance in the target coherent condition and

the impaired performance in the masker coherent condition arises. What therefore might be the

mechanism through which visual coherence with a to-be-ignored sound yields a processing

advantage?

At a cellular level, successful stream segregation is thought to be a consequence of the activation

of distinct neural populations in time, such that neurons representing the same stream are highly

temporally coherent in their responses and those representing different streams share low

coherence (Lu et al., 2017; Middlebrooks & Bremen, 2013). Under such a model temporal

coherence with either the target *or* the masker stream should result in more distinct (and hence

better segregated) neural responses that in turn offer a more effective substrate on which

selective attention can operate. If AV temporal coherence allows the representation of each of

two competing sounds to be more distinct within sensory cortex then temporal coherence

between target or masker stream should offer an advantage over an independently modulated

visual stimulus. The data from the AVTC group suggest that after training listeners were able to

benefit from AV temporal coherence when the visual stimulus was temporally coherent with

either auditory stream. Possible mechanistic explanations for this would be that training has

resulted in listeners being better able to use top-down control to actively suppress the masker

stream in the masker coherent case, which in turn enables them to better detect the timbre

deviants in the target stream. This suggestion is supported by the data in figure 4, which show a

condition-specific increase in hit-rates for the masker-coherent condition after training. While we

can only speculate about the mechanism underlying the effects observed here, a recent study that

trained listeners to improve their audiovisual temporal perception reported enhanced beta band

activity after training and suggested that enhanced top-down modulation was responsible for

improved temporal processing (Theves, Chan, Naumer, & Kaiser, 2020). Further assessment of

this hypothesis requires neurophysiological work to determine how selective attention and

audiovisual object formation interact to shape the responses to target and masker streams in

auditory cortex.

We implemented an adaptive training procedure with feedback to force subjects to work close to

their perceptual threshold for most of the training periods. In keeping with other studies (Sürig et

al., 2018), adaptive training was highly effective at rapidly driving learning in both of the

temporal discrimination tasks. While many adaptive tasks show that the majority of learning

occurs in the first session, repeated learning is thought to be critical for stabilizing learning

(Shibata et al., 2017a, 2017b). Follow up studies would be required to determine the optimal

training strategy for maximizing long term perceptual gains.

Training listeners to narrow their temporal binding window often decreases their likelihood of

integrating auditory and visual stimuli (McGovern et al., 2016; Setti et al., 2014), and the

temporal binding window itself is task and stimulus dependent (M. A. De Niear, Gupta, Baum, &

Wallace, 2018; Megevand et al., 2013; R. A. Stevenson & Wallace, 2013). Decreased integration

can be successfully modelled within a Bayesian causal inference framework as resulting from

both an increase in the precision of timing estimates and a decrease in a prior belief that signals

originate from the same source (McGovern et al., 2016). Here, we report enhanced auditory

visual integration after training listeners to make temporal coherence judgments. Unlike studies

that train listeners to narrow their perceptual binding window, in our training paradigm

participants were effectively trying to detect small amounts of temporal coherence and

distinguish this from fully independent stimuli.

The width of the temporal binding window predicts susceptibility to sound induced flash

illusions in naïve listeners (Stevenson, Fister, Barnett, Nidiffer, & Wallace, 2012). We did not

find a relationship between the ability of naïve listeners to assess temporal coherence and their

ability to exploit temporal coherence between the target and the visual stimulus. Nonetheless, we

did observe a correlation between the ability of listeners to discriminate temporal coherence and

the relative pattern of performance of the masker-coherent and independent condition with those

people who were best able to assess temporal coherence showing a benefit of masker-coherent

over independent, and those people who were worse as assessing temporal coherence being

relatively impaired on the masker coherent condition relative to the independent condition.

Previous studies have illustrated that visual cues can assist speech processing in noise (Grant,

Walden, & Seitz, 1998; Helfer & Freyman, 2005; Schwartz, Berthommier, & Savariaux, 2004).

While speech reading abilities are strongly predictive of audiovisual benefit for speech reception

thresholds (MacLeod & Summerfield, 1987), lip reading can influence auditory streaming

(Devergie, Grimault, Gaudrain, Healy, & Berthommier, 2011), supporting the idea that, in

addition to conveying phonetic information, lip reading benefits in noise potentially comprise of

both bottom-up sensory effects that facilitate auditory scene analysis (Atilgan et al., 2018)

Previous studies exploring the transfer of effects from training on temporal simultaneity

judgments to other multisensory paradigms have had mixed results with transfer occurring to

some tasks but not others (McGovern et al., 2016; Powers Iii, Hillock-Dunn, & Wallace, 2016;

Setti et al., 2014; Sürig et al., 2018). An important question in interpreting the significance of our

findings is whether the benefits in the auditory selective attention task transfer to other more

real-world tasks such as utilizing speech reading in noisy listening conditions.

Context

We have previously (Maddox et al., 2015) demonstrated that a temporally coherent visual

stimulus can enhance the ability of listeners to focus on one sound in a mixture. Using the same

stimuli, we demonstrated a bottom up mechanism through which visual information could

change the way in which sound mixtures were represented in auditory cortex, such that the

neural representation of sounds that were temporally coherent with a visual stimulus were

enhanced. One observation we made from our behavioural data was that listeners varied greatly

in their ability to use visual to augment auditory scene analysis. In this study is therefore a first

attempt to understand whether we could train listeners to use visual information more effectively.

Our longer term goal is to relate these findings to other situations that require focusing on one

sound in a mixture, such as listening to speech in noise, to understand whether listeners might

benefit from training in order to exploit visual information more effectively.

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