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How we pay attention in naturalistic settings

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Abstract

Traditional research on attentional control has largely focused on single senses and the

importance of one's behavioural goals in controlling attentional selection, thus limiting its

generalizability to real-world contexts. These contexts are both inherently multisensory and

contain regularities that also contribute to attentional control. To get a better

understanding of how attention is controlled in the real world, we investigated how visual

attentional capture was impacted by top-down goals (indexed by task-set contingent

attentional capture) and the multisensory nature of stimuli, as well as top-down contextual

factors such as semantic relationships and temporal predictability of stimulus onset.

Participants performed a multisensory version of Folk et al. (1992) spatial cueing paradigm,

while their 129-channel event-related potentials (ERPs) were recorded. Reaction-time

spatial cueing served as a behavioural measure of attentional control, while the N2pc ERP

component was analysed both canonically and using a multivariate electrical neuroimaging

(EN) framework. Behaviourally, target-congruent colour distractors captured attention more

strongly when they were simultaneous than semantically congruent (nontarget-congruent

colour distractors failed to capture attention), with no behavioural evidence for context

modulating multisensory enhancements of capture. However, our EN analyses revealed

context-based influences on attention to both visual and multisensory distractors, on how

strongly they activated brain networks and in the type of activated brain networks. In both

cases, these context-driven brain response modulations occurred early on (long before the

traditional N2pc time-window), with network-based modulations at ~30ms post-distractor,

followed by strength-based modulations at ~100ms post-distractor. Our findings revealed

that in naturalistic settings, meaning, next to predictions (spatial, temporal etc.) might be a

second important source of contextual information utilised to facilitate goal-directed

attention. Therein, attentional selection is controlled by an interaction of one's goals,

stimulus' perceptual (multisensory-driven) salience *and* an interaction of stimulus' meaning

and its predictability. Our study demonstrates how investigating more traditional, lab-

studied control mechanisms and processes more typical for everyday life reveals a complex

interplay between goal-, stimulus- and context-based processes in attentional control. As

such, our findings call for a revision of traditional models of visual attentional control to

account for the role of both contextual and multisensory control mechanisms.

Keywords: attentional control, multisensory, real-world, semantic congruence, temporal

predictability, context

                                    Introduction

Goal-directed behaviour depends on the ability to allocate processing resources towards

those stimuli that are important to the current behavioural goals of the individual – a set of

skills jointly known as "attentional control". On the one hand, our current knowledge about

attentional control may be limited to the rigorous yet highly artificial conditions in which it

is traditionally studied. On the other hand, findings from studies that assess attentional

control with naturalistic stimuli (audio recordings, films, etc.) may be limited in their validity

by confounds from other processes present in fully naturalistic settings. Here, we

systematically tested how traditionally studied goal- and salience-based attentional control

mechanisms interact with more naturalistic, context-based mechanisms using a 'naturalistic

laboratory' approach that combines the rigour of lab-based studies with variabilities present

in everyday situations (Matusz et al., 2019a).

80   In the real world, the location of goal-relevant information is rarely known in

advance. For example, when searching for a friend on a crowded city street, we may know

what our 'target' looks like, but not necessarily where it is located. For such cases, the visual

search paradigm (Treisman & Gelade, 1980) has provided a model for how attentional

control operates; it demonstrated that a target's stimulus attributes can be effective in

controlling attentional selection of objects in multi-stimulus settings. Subsequent research

scrutinised whether stimulus-based or goal-driven mechanisms have primacy in controlling

attention. While some have argued for the priority of bottom-up stimulus salience (e.g.

Theeuwes, 1991), others have proposed that the individual's current behavioural goals

guide attentional selection (Folk et al., 1992). Folk and colleagues (1992) employed a version

of spatial cueing paradigm, where participants searched for a colour-defined target in a

search array and the search array was always preceded by a distractor array, where one

element was either a colour singleton or an abrupt onset. In this colour-defined search,

attentional capture - indexed by spatial cueing (a difference in response speed to targets in

cued versus uncued locations) - was elicited only by colour-defined distractors. This

prompted them to propose the 'task-set contingent attentional capture' (or TAC)

hypothesis, whereby salient but irrelevant objects will only capture attention if they share

features with the target and are therefore potentially task-relevant. In other words, top-

down goal-based factors had primacy in controlling attentional selection.

However, it has quickly become clear that to understand how we pay attention in
the real world, we must expand study designs to include components of real-world
situations, such as the spatiotemporal or semantic information contained in the stimulus
and/or the environment within which it occurs (e.g., Chun & Jiang 1998; Peelen & Kastner,
2014; Summerfield et al., 2006; van Moorselaar & Slagter 2019; Press et al. 2020) as well as
the multisensory nature of objects in those environments (Matusz & Eimer, 2011, 2013;
Matusz et al. 2015; Lunn et al. 2019; Soto-Faraco et al. 2019). One step towards increasing
the ecological validity of the dominating models of attentional control is to systematically
test how visually-derived goal- and context-based mechanisms interact with processes that
are based on integration of information across the senses (henceforth, multisensory
processes). The integration of simultaneous inputs across multiple senses, has been
observed within the first 100ms post-stimulus onset, both in humans, as well as alert and
even anaesthetized animals, and regardless of task-relevance (e.g., Fort et al. 2002;
Ghazanfar et al. 2005; Giard and Peronnet 1999; Kayser et al. 2008; Rowland & Stein, 2007;
Talsma & Woldroff, 2005; reviewed in de Meo et al., 2015). The early latency and putative
source in low-level cortices (e.g., Cappe et al., 2010; Murray et al., 2016a; Raij et al., 2010)
of such multisensory integration based on pure simultaneity detection suggests that
multisensory processes may influence attentional selection in a bottom-up fashion, perhaps
even independently of top-down goal-based attentional control (Matusz & Eimer, 2011).
Matusz and Eimer (2011) adapted the Folk et al.'s (1992) spatial cueing paradigm to
multisensory settings and demonstrated that indeed visual distractors capture attention
when they match target's feature, such as colour, replicating the TAC effect. However, they
also showed the same visual distractors captured attention more strongly when
accompanied by a sound, regardless of their goal-relevance. This study revealed the
importance of such bottom-up multisensory enhancement (MSE) for attentional selection of
visual objects even under strong top-down goal-based visual control. However, relevance to
behavioural goals is just one of many top-down attentional control mechanisms impacting
behaviour in the outside world. Many such mechanisms have been studied in the visual
domain (Summerfield & Egner, 2009; Gazzaley & Nobre, 2012; Peelen & Kastner, 2014). Two

such factors that deserve special attention are the temporal and semantic organisation of
stimuli within a given 'context'[1].

Top-down contextual factors of attentional control
Temporal structure of the environment is routinely used to build predictions. In attentional
contexts, attentional control will utilise those predictions to improve selection of target
stimuli (e.g., Correa et al., 2005; Coull et al., 2000; Green & McDonald, 2010; Miniussi et al.,
1999; Naccache et al., 2002; Rohenkohl et al., 2014). Predictions are likewise utilised in
service of inhibition of task-irrelevant stimuli albeit in that area the role of location or
feature-based predictions is more researched (reviewed in Noonan et al. 2018; van
Moorselaar & Slagter 2020a). In naturalistic, multisensory settings, temporal predictions
have been shown to be particularly important, where they aid, e.g., decoding meaning from
syllables (ten Oever & Sack, 2015), sentences (Luo & Poeppel, 2007) and across competing
streams of information (Zion Golumbic, Poeppel, & Schroeder, 2012). What remains
unknown is whether temporal predictions in a similar vein facilitate top-down goal-based
attentional control in real-world multisensory settings. In one recent example, we
demonstrated how spatial predictability can be effectively utilised to suppress the
processing of unisensory goal-irrelevant auditory distractors (Matusz et al. 2016).

Apart from low-level stimulus regularities like temporal predictability, semantic
relationships embedded in real-world environments are another basic organising principle
of information in such contexts (Amedi et al., 2005; Murray et al., 2016b; ten Oever et al.
2016; Soto-Faraco et al., 2019). There has been plentiful evidence that semantic
relationships influence the processing of multisensory objects, in that semantically
congruent audiovisual stimuli are more effectively identified and searched, compared to
over incongruent stimuli (e.g., Chen & Spence, 2010; Iordanescu et al., 2008; Tovar et al.
2020; for review see Doehrmann & Naumer, 2008). Since real-world contexts are composed
of meaningful objects, the brain likely has adapted to benefit from the redundancy of
semantic information within multisensory objects in external environments, perhaps even
more so than from redundancies in spatio-temporal information (Alais et al. 2010; ten

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[1] Context has been previously defined as the "immediate situation in which the brain operates... shaped by external circumstances, such as properties of sensory events, and internal factors, such as behavioural goal, motor plan, and past experiences" (van Atteveldt et al., 2014).

Oever et al. 2016). For example, Murray and colleagues have showed that semantically

congruent multisensory pairings are better remembered than arbitrary (simultaneous)

multisensory pairings (Murray et al., 2004; 2005; Matusz et al., 2015; Thelen et al. 2015,

reviewed in Matusz, Wallace & Murray, 2017, 2020). It remains unclear, however, if

semantic congruence would likewise enhance attentional control over multisensory stimuli

over and above the enhancement engendered by multisensory simultaneity.

Taken together, the evidence suggests that predictions about stimulus onset and

meaning generated through experience with the environment are commonly employed to

facilitate achieving one's current behaviour goals. In line with Bayesian frameworks, the

brain utilises such to bias interpretation of sensory information (and action) *towards* those

predictions (e.g. Summerfield & Egner, 2009). Alternatively, the brain is thought to suppress

expected inputs on behalf of processing of the most informative, "surprising" events (e.g.

Richter et al. 2018) or do so when the stimulus-driven surprise / informativeness is

particularly high (e.g. Press et al. 2020). Notwithstanding, the interplay between

simultaneously occurring behavioural goals, multisensorily-generated perceptual salience

and spatiotemporal/ semantic predictions in real-world multisensory settings remains

practically unresearched (see Ten Oever et al., 2016; Matusz et al. 2019a, 2019b). In order

to ascertain the various, potentially interacting, influences that could affect attentional

control in multisensory settings, the present study employed an experimental paradigm that

emulates real-world conditions while retaining a sufficient degree of experimental control

to systematically manipulate all the factors of interest.

The present study

In the present study, we investigated whether contextual factors influence attentional

control processes such as visual top-down control and bottom-up multisensory control.

Specifically, we asked whether attentional control over visual and audiovisual distractors is

influenced by the temporal predictability of their onset or the multisensory relationship

between auditory and visual distractor features.

For this purpose, we used the multisensory version of Folk's spatial cueing paradigm

developed by Matusz and Eimer (2011). This multisensory paradigm allows a direct and

systematic comparison of well-known visual control processes and lesser known

multisensory control processes. Top-down visual attentional control was behaviourally

assessed through the presence of TAC. Bottom-up control of attention by multisensory
stimuli was assessed through the presence of MSE. We were interested in whether the
predictability of the onset, on the one hand, and semantic relationships between visual and
auditory distractor features would modulate the TAC and MSE effects observed for visual
and audiovisual distractors.

The behavioural paradigm was combined with a simultaneous EEG recording. Event-
related potentials (ERPs) were time-locked to distractor onset and were analysed both with
canonical measures and using a multivariate whole-scalp electrical neuroimaging (EN)
framework. The canonical measure was the N2pc component; a negative-going voltage
deflection around 200ms post-stimulus onset at posterior electrode sites contralateral to
stimulus location (Luck & Hillyard, 1994a, 1994b; Eimer, 1996; Girelli & Luck, 1997) and a
classic ERP correlate of the attentional selection of potential visual targets (Eimer et al.,
2009; Eimer & Kiss, 2008; Kiss et al., 2008a; 2008b). The N2pc has been shown to be
sensitive to some contextual influences, especially match with semantic category (e.g. Wu
et al., 2015), with one notable study showing also prediction-based influences (Burra &
Kerzel, 2013). However, there is little evidence that N2pc is suitable to measure
behaviourally-observed bottom-up enhancements of attentional capture by multisensory
distractors (e.g. van der Burg et al. 2011). Notwithstanding, canonical N2pc analyses tend to
be limited in their neuropsychological interpretability (for a more detailed discussion, see
Methods, and Matusz et al., 2019b), which are readily overcome with the use of EN. It is for
this reason that we combined EN measures with canonical N2pc measures in an attempt to
elucidate the mechanisms of attentional selection in real world settings.

We expected to replicate the well-established TAC effect. In behaviour, this
encompassed reliable spatial cueing effects for target-congruent colour distractors, and
zero- or negative cueing, indicating suppression of capture, for nontarget-congruent
distractors (as in e.g., Folk et al., 1992; Folk, et al., 2002; Lien et al., 2008). Likewise, we
expected enhanced N2pc amplitudes for target-colour congruent visual distractors
compared to nontarget-colour congruent distractors (as in, e.g., Eimer et al., 2009). We
expected that as a reflection of the brain's strength of top-down visual attentional control,
TAC would be reduced for temporally predictable and/or semantically congruent (i.e.
embedded with meaning following the Training phase of the study) distractors. For
example, *spatially* predictable visual distractors were recently shown to be effectively

inhibited and this inhibition is modulated by feature-based predictions (e.g., Van

Moorselaar & Slagter 2019, 2020b). In turn, semantic relationships have been previously

shown to enhance visual target selection (e.g. Wu et al., 2015), indicating reliance on goal-

based mechanisms. As for MSE, we expected to replicate the behavioural enhancement

observed by Matusz and Eimer (2011), similarly expecting that at least one of the contextual

factors would affect MSE, where we expected it to reduce MSE. For example, predictability

of location of sound distractors is known to eliminate sound-induced activations of occipital

responses (Matusz et al. 2016). In turn, similarly to visual symbols, also multisensory

semantic relationships are known to enhance visual target selection across the senses (e.g.

Iordanescu et al., 2008).

We did not have strong predictions whether the interactions with context-based

processes on behavioural capture effects will be accompanied by concomitant N2pc

modulations. For TAC, known to be well-reflected in N2pc (e.g. Eimer et al. 2009), we

predicted it may be further modulated by temporal predictability (Burra & Kerzel, 2013) or

semantic relationships (Wu et al. 2015) albeit existing research focused on target-elicited

N2pc's and did not measure context-based modulations over the difference between target-

colour vs. nontarget-colour N2pcs; this leaves the question of context-based modulations of

TAC on distractor-elicited N2pc's open. With respect to MSE, there is little evidence for

sensitivity of N2pc to perceptual salience, visual or otherwise, of distractors (cf. Eimer et al.

2009; van der Burg et al. 2011), thus we did not have strong predictions whether distractor-

elicited N2pc's would be stronger for audiovisual than visual distractors, or whether this

MSE would interact with context-based processes. We likewise had no strong expectations

as to whether the two contextual factors would interact in modulating, either TAC or MSE.

Finally, with limited prior knowledge on the potential contextual influences on TAC

and MSE, we likewise investigated the role of lateralised, N2pc-like and nonlateralised

mechanisms in supporting the interactions between TAC and MSE on the one hand, and

contextual factors on the other hand. As before (Matusz et al. 2019b), we have used the

well-researched N2pc component to guide our EN analyses of the contextual influences over

TAC and MSE. That is, in EN, we analysed lateralised ERPs at ~180–300ms after stimulus

onset that are captured by the canonical N2pc component analyses, to study the

importance of strength- and network-based lateral brain mechanisms in supporting the

studied interactions. However, given the exploratory nature of these EN analyses, and

previous research that has demonstrated nonlateralised ERP modulations by semantic

information and temporal expectations (e.g., Dell'Acqua et al., 2010; Dassanayake et al.,

2016), we also analysed nonlateralised ERPs within an EN framework and did so across the

whole post-stimulus time-period.

258   **Materials and Methods**

Participants

Thirty-nine adult volunteers participated in the study (5 left-handed, 14 male, M_{age}:

27.5years, SD: 4years, range: 22–38years). Participants had normal or corrected-to-normal

vision and normal hearing and reported no prior or current neurological or psychiatric

disorders. Participants provided informed consent before the start of the experiment. All

research procedures were approved by the Cantonal Commission for the Ethics of Human

Research (CER-VD; no. 2018-00241).

Task properties and procedures

Each participant took part in one testing session consisting of four experimental tasks

(henceforth referred to as 'experiments', Figure 1A), where the first two experiments were

followed by a training task (henceforth referred to as 'training', Figure 1C). The first two

tasks involved non-semantically related colour-pitch combinations as in the original study,

while the last two involved colour-pitch combinations that were semantically congruent (the

factor of Multisensory Relationship in Figure 1B). Such a semantic relationship between

colours and sounds was created using a training task that was based on a semantic

association task (described in detail below; Sui et al. 2012; see also Sun et al., 2016).

Further, Experiments 1 and 3 involved distractor onsets variable in duration, while

Experiments 2 and 4 involved cues that had a constant onset (the factor of Distractor Onset

in Figure 1B). As a pilot study revealed sufficient proficiency at the experimental task (over

50% accuracy) after a few trials, participants did not practice the task before its

administration.

Experimental and training tasks were conducted in a dimly lit, sound-attenuated

room, with participants seated at a distance of 90 cm from a 23" LCD monitor with a

resolution of 1080 × 1024 (60-Hz refresh rate, HP EliteDisplay E232). All visual elements

were approximately equiluminant ($\sim$20cd/m^2), as determined by a luxmeter (model SC160,

CESVA Instruments) placed at a position adjacent to participants' eyes, measuring the

luminance of the screen filled with each respective element's colour. The averages of three

measurement values per colour were averaged across colours and transformed from lux to

cd/m^2 in order to facilitate comparison with the results of Matusz & Eimer (2011). The

testing session lasted no longer than 3h in total, including an initial explanation and

obtaining consent, EEG setup, experiments and training, and breaks.

Experiments. Across the experimental tasks, participants searched for a colour

predefined target (e.g., a red bar) in a search array, and assessed the target's orientation

(vertical vs. horizontal). The search array was always preceded by an array containing

distractors. Distractors were visual and audiovisual stimuli that could match the target

colour (red set of dots) or not match the target colour (blue set of dots), as in the original

Matusz and Eimer (2011) study (Exp.2).

Like in the former study, each experimental trial consisted of the following sequence

of arrays: base array, followed by distractor array, followed by a fixation point, and finally a

target array (Figure 1A). However, compared to the original Matusz and Eimer (2011)

paradigm, the number of elements was reduced from 6 to 4 and targets were reshaped to

look like diamonds rather than rectangles, as Experiment 1 served as an adult control for a

different, developmental study (Turoman et al., in prep.). Thus, the base array contained

four differently coloured sets of closely aligned dots, each dot subtending 0.1° × 0.1° of

visual angle. Elements were spread equidistally along the circumference of an imaginary

circle against a black background, at an angular distance of 2.1° from a central fixation point.

Each set element could be one of four possible colours (according to the RGB scale): green

(0/179/0), pink (168/51/166), gold (150/134/10), silver (136/136/132). In the distractor

array, one of the base array elements changed colour to either a target-congruent colour, or

a nontarget-congruent colour that was not present in any of the elements before. The

remaining three distractor array elements did not change their colour. The cues and the

subsequent target "diamonds" could have either a blue (RGB values: 31/118/220) or red

(RGB values: 224/71/52) colour. The target array contained four bars (rectangles) where one

was always the colour-defined target. Target colour was counterbalanced across

participants. Target orientation (horizontal or vertical) was randomly determined on each

trial. The two distractor colours were randomly selected with equal probability before each

trial, and the colour change was not spatially predictive of the subsequent target location

(same distractor– target location on 25% of trials). On half of all trials, distractor onset

coincided with the onset of a pure sine-wave tone, presented from two loudspeakers on the

left and right sides of the monitor. Sound intensity was 80 dB SPL, as measured using an

audiometer placed at a position adjacent to participants' ears.

Manipulations of distractors' target-colour congruent and the presence/absence of

sound resulted in four general distractor conditions: TCCV (target-congruent colour

distractor visual), NCCV (nontarget-congruent colour, Visual), TCCAV (target-congruent

colour distractor, AudioVisual), NCCAV (nontarget-congruent colour, AudioVisual). These

conditions translated into 3 factors that were comparable to Matusz and Eimer's original

design: Distractor Colour (target-congruent colour - TCC vs. nontarget-congruent colour -

NCC), Distractor Modality (Visual - V vs. AudioVisual - AV) and Cue-Target Location (Same vs.

Different). Further manipulations included the introduction of two contextual factors:

Distractor Onset and Multisensory Relationship, which were manipulated across the four

experimental tasks. For the Distractor Onset factor, in Experiments 2 and 4, base array

duration (and therefore distractor onset) was kept constant at 450ms, as in the original

Matusz and Eimer (2011) paradigm. Meanwhile in Experiments 1 and 3, base array duration

(and therefore distractor onset) was varied between 100, 250 and 450ms. This way, the

strength of attentional capture by temporally predictable distractors could be compared

with the capture elicited by unpredictable distractors, and if and how these visual and

audiovisual distractors differ on that dimension. For the Multisensory Relationship factor,

the sound frequency was set to 2000Hz in Experiments 1 and 2, as in the Matusz and Eimer

(2011) paradigm, and alternated between 4000Hz (high-pitch) or 300Hz (low-pitch) in

Experiments 3 and 4. Then, a Training was presented after Experiment 2, in order to induce

in participants a semantic-level association between a specific distractor colour and a

specific pitch (Figure 1C). Specifically, following the Training, on AV trials, tones were always

congruent ('matched') with the colour of the cue that they were presented with. (e.g.,

participants who were taught to associate red and high-pitch sounds and blue colour and

low-pitch sounds would, after the Training, only be presented with those AV pairings). This

way, the strength of attentional capture by colour-pitch combinations characterised only by

their simultaneous presentation could be compared with the capture elicited by colour-

pitch combinations characterised by semantic congruence.

To summarise, the study design included 5 different factors. These were Distractor
Colour, Distractor Modality, and Cue-Target Location, and two new factors: Distractor Onset
(DO; Predictable vs. Unpredictable) and Multisensory Relationship (MR; Simultaneity vs.
Congruence). The full experimental session consisted of 8 blocks of 64 trials each, for each
of the 4 experiments, resulting in 2,048 trials in total (512 trials per experiment).
Participants were told to respond as quickly and accurately as possible to the targets'
orientation by pressing one of two horizontally aligned round buttons (Lib Switch, Liberator
Ltd.) that were fixed onto a tray bag on the participants' lap. If participants did not respond
within 5000ms of the target presentation, the next trial was initiated, otherwise the next
trial was initiated immediately after a button press. Feedback on accuracy was given after
each block, followed by a 'progress (treasure) map' which informed participants of the
number of blocks remaining until the end, and during which participants could take a break.
Breaks were also taken between each experimental task.

Training. The Training procedure consisted of an Association phase and a Testing
phase. In the Association phase, participants were shown alternating colour word–pitch
pairs. Specifically, each pair consisted of a word, denoting a distractor colour, that was
presented on the centre of the screen at the same time as a spatially diffuse pure tone that
was either high (4000Hz) or low (300Hz) in pitch. Both the colour word and sound were
presented for 2 seconds, after which a central fixation cross was presented for 150ms,
followed by the next colour word-pitch pair. Colour words were paired with sounds
according to two possible pairing options. In one pairing option, the high-pitch tone was
associated with the word 'red' and the low-pitch tone with the word 'blue', and in another
pairing option, the high-pitch tone was associated with the word 'blue' and the low-pitch
tone with the word 'red' (Figure 1C, Association phase). Pairing options were
counterbalanced across participants. Therefore, if the first pairing option was selected, a
presentation of the word 'red' with a high-pitch tone would be followed by a presentation
of the word 'blue' with a low-pitch tone, which would again be followed by the former pair,
etc. There were ten presentations per pair, resulting in a total of 20 trials. Colour words
were chosen instead of actual colours to ensure that associations were based on semantic
meaning rather than a linking of basic stimulus features (for examples of such taught
crossmodal correspondences see e.g., Ernst, 2007). Colour words were shown in
participants' native language (speakers: 19 French, 8 Italian, 5 German, 4 Spanish, 3 English).

Participants were instructed to observe and try to memorise the pairings as best as they
could, as they would be subsequently tested on how well they learnt the pairings.

The strength of colour-pitch associations was assessed in the Testing phase. Here,
participants were shown colour word-pitch pairings (as in the training) as well as colour-
pitch pairings (a string of x's in either red or blue paired with a sound, Figure 1C, Testing
phase). Based on the pairing option that participants were 'taught' in the Association
phase, pairings could be either matched or mismatched. For example, if 'red' was paired
with a high-pitch tone in the Association phase, in the Testing phase, the word 'red' (or red
x's) paired with a high-pitch tone would match, while the word 'red' (or red x's) paired with
a low-pitch tone would be mismatched. Participants had to indicate whether a given pair
was matched or mismatched by pressing an appropriate button on the same response setup
as in the experiments. In a similar paradigm used by Sui et al. (2012), people were able to
reliably associate low-level visual features (colours and geometric shapes) with abstract
social concepts such as themselves, their friend, and a stranger. Following their design, in
the Testing phase each pairing was shown for 250ms, of which 50ms was the sound (instead
of the stimulus duration of 100ms that Sui et al. (2012) used, to fit our stimulus
parameters), followed by an 800ms blank screen where choices were to be made, and
feedback on performance after each answer was given. Before each trial, a fixation cross
was shown for 500ms. Each participant performed three blocks of 80 trials, with 60 trials
per possible combination (colour word – sound matching, colour word – sound
nonmatching, colour – sound matching, colour – sound nonmatching). A final summary of
correct, incorrect, and missed trials was shown at the end of testing phase.

EEG acquisition and preprocessing
Continuous EEG data sampled at 1000Hz was recorded using a 129-channel HydroCel
Geodesic Sensor Net connected to a NetStation amplifier (Net Amps 400; Electrical
Geodesics Inc., Eugene, OR, USA). Electrode impedances were kept below 50kΩ, and
electrodes were referenced online to Cz. First, offline filtering involved a 0.1Hz high-pass
and 40Hz low-pass as well as 50Hz notch and a second-order Butterworth filter (–
12dB/octave roll-off, computed linearly with forward and backward passes to eliminate
phase-shift). Next, the EEG was segmented into peri-stimulus epochs from 100ms before
distractor onset to 500ms after distractor onset.

A) General experimental trial sequence

B) Overall structure of the study

Distractor Predictability

Multisensory Relationship

① simultaneous unpredictable

② simultaneous predictable

Training

④ congruent predictable

③ congruent unpredictable

C) Training of semantic audio-visual associations for distractors

Figure 1. A) An example trial of the General experimental task is shown, with four successive arrays. The white circle around the target location (here the target is a blue diamond) and the corresponding distractor location serves to highlight, in this case, a nontarget-colour distractor condition, with a concomitant sound, i.e., NCCAV. **B)** The order of tasks, with the corresponding conditions of Multisensory Relationship (MR) in red, and Distractor Onset (DO) in green, shown

separately for each experiment. Predictable and unpredictable blocks before and after the training
(1 & 2 and 3 & 4, respectively) were counterbalanced across participants. C) Events that were part of
the Training. *Association phase*: an example pairing option (red – high pitch, blue – low pitch) with
trial progression is shown. *Testing phase*: the pairing learnt in the Association phase would be tested
using a colour word or a string of x's in the respective colour. Participants had to indicate whether
the pairing was correct via a button press, after which feedback was given. Following the Training
participants always encountered 'semantically congruent' (associated during the Training)
audiovisual pairings, e.g. red – high-pitch and blue – low-pitch AV pairings.

An automatic artefact rejection criterion of ±100μV was used, along with visual inspection.
Epochs were then screened for transient noise, eye movements, and muscle artefacts using
a semi-automated artefact rejection procedure. Data from artefact contaminated electrodes
were interpolated using three-dimensional splines (Perrin et al., 1987). Across all
experiment, 11% of epochs were removed on average and 8 electrodes were interpolated
per participant (6% of the total electrode montage).

Cleaned epochs were averaged, baseline corrected to the 100ms pre-distractor time
interval, and re-referenced to the average reference. To eliminate residual environmental
noise in the data, 50Hz filter was applied. All of the above steps were done separately for
ERPs from the four distractor conditions, separately for cues in the left and right hemifield.
To analyse cue-elicited lateralised ERPs, single-trial data from all conditions with cues
presented on the left were relabelled to have electrodes over the left hemiscalp represent
activity over the right hemiscalp, and vice versa. After relabelling, the "mirror cue-on-the-
right" single-trial data and the veridical "cue-on-the-right" data were averaged together,
creating a single average ERP for each of the 4 distractor conditions. This way, the
contralaterality that would normally be represented by separate ERPs, was preserved within
each averaged ERP, with contralateral and ipsilateral hemiscalp activations present within
each averaged ERP. As a result, we obtained 4 different ERPs: TCCV (target colour-cue,
Visual), NCCV (nontarget colour-cue, Visual), TCCAV (target colour-cue, AudioVisual), NCCAV
(nontarget colour-cue, AudioVisual). Preprocessing and EEG analyses, unless otherwise
stated, were conducted using CarTool software (available for free at
www.fbmlab.com/cartool-software/; Brunet, Murray, & Michel, 2011).

Data analysis design

Behavioural analyses. Like in Matusz and Eimer (2011), and because mean reaction
times (RTs) and accuracy did not differ significantly between the four experiments, the basis
of our analyses were RT spatial cueing effects (henceforth "behavioural capture effects").
These were calculated by subtracting the mean RTs for trials where the distractor and target
were in the same location from the mean RTs for trials where the distractor and the target
location differed, separately for each of the four distractor conditions. RT data were
normally distributed and consequently were analysed using the repeated-measures analysis
of variance (rmANOVA). Error rates (%) were also analysed. As they were not normally
distributed, we analysed error rates using the Kruskal–Wallis H test and the Durbin test. The
former was used to analyse if error rates differed significantly between experiments, while
the latter was used to analyse differences between experimental conditions within each
experiment separately.

Following Matusz and Eimer (2011), RT data were cleaned by discarding incorrect
and missed trials, as well as RTs below 200ms and above 1000ms. Additionally, to enable
more direct comparisons with the developmental study for which current Experiment 1
served as an adult control (Turoman et al. 2020), we have further removed trials with RTs
outside 2.5SD of the individual mean RT. As a result, a total of 5% of trials across all
experiments were removed. Next, behavioural capture effects were submitted to a four-
way 2 × 2 × 2 × 2 rmANOVA with factors: Distractor Colour (TCC vs. NCC), Distractor
Modality (V vs. AV), Multisensory Relationship (MR; Simultaneity vs. Congruence), and
Distractor Onset (DO; Unpredictable vs. Predictable). Due to the error data not fulfilling
criteria for normality, we used Cue-Target location as a factor in the analysis, conducting 3-
way Durbin tests for each experiment, with factors Distractor Colour, Distractor Modality,
and Cue-Target Location. All analyses, including post-hoc paired t-tests, were conducted
using SPSS for Macintosh 26.0 (Armonk, NY: IBM Corp). For brevity, we only present the RT
results in the Results, and the error rate results can be found in Supplemental Online
Materials ("SOMs" henceforth).

ERP analyses. The preprocessing of the ERPs triggered by the visual and audiovisual
cues across the 4 different experimental blocks created ERP averages in which the
contralateral versus ipsilateral ERP voltage gradients across the whole scalp were preserved.
We first conducted a canonical N2pc analysis, as the N2pc is a well-studied and well-

understood correlate of attentional selection in visual settings. However, it is unclear if the

N2pc also indexes bottom-up attentional selection modulations by multisensory stimuli, or

top-down modulations by contextual factors like multisensory semantic relationships (for

visual-only study, see e.g., Wu et al. 2015) or stimulus onset predictability (for visual-only

study, see e.g., Burra & Kerzel, 2013). N2pc analyses served also to bridge EN analyses with

the existing literature and EEG approaches more commonly used to investigate attentional

control. Briefly, EN encompasses a set of multivariate, reference-independent analyses of

global features of the electric field measured at the scalp (Biasiucci et al., 2019; Koenig et

al., 2014; Michel & Murray, 2012; Murray, Brunet, & Michel, 2008; Lehmann & Skrandies,

1980; Tivadar & Murray, 2019; Tzovara et al., 2012). Below we describe only the analyses

conducted as part of the present study. EN analyses offer two key advantages over

canonical N2pc analyses. First, they consider signal from the entire electrode montage,

whereas N2pc analyses only do so for two electrodes in a montage, potentially overlooking

relevant modulations occurring in other areas of the scalp electric field. Second, N2pc

analyses can only reveal average amplitude differences between experimental conditions,

but not the brain mechanisms underlying these differences. In contrast, EN analyses can

reveal whether brain response modulations between experimental conditions arise from: 1)

differences in response strength within a single brain network, or 2) differential brain

networks being active for different conditions. We focused our analyses on EN analyses of

lateralised ERPs (which are the centre of the canonical N2pc analyses), but instead of using a

pair of contralateral channels, we analysed modulations across the entire 129-channel

montage.

Canonical N2pc analysis. To analyse N2pc in traditional way, we extracted mean

amplitude values from, first, two electrode clusters comprising PO7/8 electrode equivalents

(e65/90; most frequent electrode pair used to analyse the N2pc), and, second, their six

immediate surrounding neighbours (e58/e96, e59/e91, e64/e95, e66/e84, e69/e89,

e70/e83), over the 180–300ms post-distractor time-window. Analyses were conducted on

the mean amplitude of the N2pc difference waveforms, which were obtained by subtracting

the average of amplitudes in the ipsilateral posterior-occipital cluster from the average of

amplitudes in the contralateral posterior-occipital cluster. This step helped mitigate the loss

of statistical power that could result from the addition of contextual factors into the design.

N2pc means were thus submitted to a 4-way 2 × 2 × 2 × 2 rmANOVA with factors Distractor

Colour (TCC vs. NCC), Distractor Modality (V vs. AV), MR (Simultaneity vs. Congruence), and
Distractor Onset (DO; Unpredictable vs. Predictable), analogously to the behavioural
analysis.

Electrical Neuroimaging of the N2pc component. EN analyses separately tested
response strength and topography (see e.g. Matusz et al., 2019b for a tutorial-like
description of how EN measures can be applied to study attentional control processes). We
assessed whether top-down and bottom-up factors modulated the cue-elicited lateralised
ERPs by altering the strength of responses within statistically indistinguishable brain
networks. For this purpose, we analysed Global Field Power (GFP), which is the root mean
square of potential [µV] across the entire electrode montage (see Lehmann & Skrandies,
1980).

Subsequently, we analysed stable patterns of ERP topography that characterise
experimental conditions using a clustering approach known as the Topographic Atomize and
Agglomerate Hierarchical Clustering (TAAHC). This clustering (also known as
"segmentation") procedure generates sets of clusters of topographical maps that predict
the largest variance within the group-averaged ERP data. Each cluster is labelled with a
'template map' that represents the centroid of its cluster. The optimal number of clusters is
one that explains the largest global explained variance (GEV) in the group-averaged ERP
data with the smallest number of template maps, and which we identified using the
modified Krzanowski–Lai criterion (Murray et al., 2008). In the next step, i.e., the so-called
fitting procedure, the single-subject data was 'fitted' back onto the segmentation results,
such that each datapoint of each subject's ERP data over a chosen time-window was
labelled by the template map with which it was best spatially correlated. This procedure
resulted in a number of timeframes that a given template map was present over a given
time-window, which durations (in milliseconds) we then submitted to statistical analyses
described below. More generally, we hasten to remind the reader that if two ERPs differ in
topography, they also differ in the configurations of underlying brain sources (Lehmann,
1987; Vaughan Jr, 1982).

In the present study, we conducted strength-based and topography-based analyses
using the same 4-way repeated measures design as in the behavioural and N2pc analyses,
on both lateralised and nonlateralised whole-montage ERP data. Since the N2pc is a
lateralised ERP, we first conducted an EN analysis of lateralised ERPs in order to uncover the

modulations of the N2pc by contextual factors. Thus, in order to obtain *global* EN measures
of *lateralised* N2pc effects, we computed a difference ERP by subtracting the voltages over
the contralateral and ipsilateral hemiscalp, separately for each of the 4 distractor
conditions. This resulted in a 59-channel difference ERP (as the midline electrodes from the
129-electrode montage were not informative). Next, this difference ERP was mirrored onto
the other side of the scalp, recreating a "fake" 129 montage (with values on midline
electrodes now set to 0). It was on these mirrored "fake" 129-channel lateralised difference
ERPs that lateralised strength-based and topography-based EN analyses were performed.
Here, GFP was extracted over the canonical 180–300ms N2pc time-window and submitted
to a 2 × 2 × 2 × 2 rmANOVA with factors Distractor Colour (TCC vs. NCC), Distractor Modality
(V vs. AV), as well as the two new factors, MR (Simultaneity vs. Congruence), and Distractor
Onset (DO; Unpredictable vs. Predictable). Meanwhile, for topographic analyses, the "fake"
129-channel data across the 4 experiments were submitted to a segmentation over the
entire post-distractor period. Next, the data were fitted back over the 180–300ms period.
Finally, the resulting number of timeframes (in ms) was submitted to the same rmANOVA as
the GFP data above.

Even though the N2pc is lateralised, it remains unknown whether the employed
contextual factors impact the N2pc at all. Given evidence that semantic information and
temporal expectations can modulate *nonlateralized* ERPs within the first 100-150ms post-
stimulus (e.g., Dell'Acqua et al., 2010; Dassanayake et al., 2016), we also investigated the
influence of contextual factors on nonlateralized voltage gradients, in an exploratory
fashion. It must be noted that ERPs are sensitive to the inherent physical differences in
visual and audiovisual conditions. Specifically, on audiovisual trials, the cue-induced ERPs
would be contaminated by brain response modulations induced by sound processing, with
these modulations visible in our data already at 40ms post-cue. Consequently, any direct
comparison of visual-only and audiovisual ERPs would index auditory processing per se and
capture of attention by audiovisual stimuli. Such confounded sound-related activity is
eliminated in the canonical N2pc analyses through the contralateral-minus-ipsilateral
subtraction. Here, to eliminate this confound in our exploratory EN analyses, we calculated
new difference ERPs, first between TCCAV and NCCAV conditions, and then between TCCV
and NCCV conditions. Such difference ERPs, just as the canonical N2pc difference waveform,
subtracted out the sound processing confound in visually-induced ERPs. These difference

ERPs produced a new factor that indexed the enhancement of visual attentional control by

sound presence that we henceforth refer to as Target Difference. All nonlateralised EN

analyses were conducted on these difference ERPs and included the factor Target Difference

with two levels: D_{AV} (TCCAV – NCCAV difference) and D_V (TCCV – NCCV difference).

As part of strength-based analyses, voltage and GFP data were analysed using the

STEN toolbox 1.0 (available for free at https://zenodo.org/record/1167723#.XS3lsi17E6h),

and submitted to 3-way rmANOVAs with factors: MR (Simultaneity vs. Congruence), DO

(Unpredictable vs. Predictable), and Target Difference (D_{AV} vs. D_V). Follow-up tests involved

further ANOVAs and pairwise *t*-tests. To correct for temporal and spatial correlation (see

Guthrie & Buchwald, 1991), we applied a temporal criterion of >15 contiguous timeframes,

and a spatial criterion of >10% of the 129- channel electrode montage at a given latency for

the detection of statistically significant effects at an alpha level of 0.05. As part of

topography-based analyses, we segmented the ERP difference data across the post-

distractor and pre-target onset period (0 – 300ms from distractor onset) and conducted

clustering of the data to obtain template maps. Next, the data were fitted onto the

canonical N2pc time-window (180–300ms), as well as time-periods that were highlighted by

the GFP data as representing significant condition differences. The resulting map presence

(in ms) over the given time-windows were submitted to 4-way rmANOVAs with factors: MR

(Simultaneity vs. Congruence), DO (Unpredictable vs. Predictable), Target Difference (D_{AV} vs.

D_V), and Map (different numbers of maps for different time-windows), followed by post-hoc

t-tests. Maps with durations <15 contiguous timeframes were not included in the analyses.

Unless otherwise stated in the Results, map durations were statistically different from 0ms

(as confirmed by post-hoc one-sample t-tests), meaning that they were reliably present

across the time-windows of interest. Holm-Bonferroni corrections (Holm, 1979) were used

to correct for multiple comparisons between map durations. Comparisons passed the

correction unless otherwise stated.

607   **Results**

Behavioural analyses

Interaction of TAC with contextual factors

Our 2 × 2 × 2 × 2 rmANOVA revealed a number of main effects and interactions, both

expected and unexpected (full description of the results in SOMs). As expected, behavioural

capture effects showed the TAC, evidenced by a significant main effect of Distractor Colour,

$F_{(1, 38)} = 340.4$, $p < 0.001$, $\eta_p^2 = 0.9$. Specifically, TCC distractors (42ms), but not NCCs (-1ms),

elicited reliable behavioural capture effects.

The strength of top-down visual attention indexed by TAC was dependent on the

multisensory relationship, demonstrated by a 2-way Distractor Colour × MR interaction, $F_{(1,}$

$_{38)} = 4.5$, $p = 0.041$, $\eta_p^2 = 0.1$ (Figure 2). This effect was driven by behavioural capture effects

elicited by TCC distractors being reliably larger under simultaneity (45ms) than congruence

(40ms), $t_{(38)} = 1.9$, $p = 0.027$. Capture effects elicited by NCC distractors showed no evidence

of a modulation by MR (Simultaneity vs. Congruence: -2ms vs. -0.2ms, $t_{(38)} = 1$, $p = 0.43$). In

contrast, there was no evidence for the strength of TAC to be dependent on predictability of

the distractor onset, with no 2-way Distractor Colour × DO interaction, $F_{(1, 38)} = 2$, $p = 0.16$.

Figure 2. Larger behavioural capture elicited by target-colour cue (TCC) distractors were found for

simultaneous than for semantically congruent distractors. Expectedly, regardless of Multisensory Relationship,

attentional capture was larger for target-colour (TCC) cues than for non-target colour cues (NCC).

Interaction of MSE with contextual factors

Behavioural capture effects likewise showed MSE, evidenced by a significant main effect of

Distractor Modality, $F_{(1, 38)}$ = 13.5, p = 0.001, η_p^2 = 0.3, where visually-elicited behavioural

capture effects (18ms) were enhanced on audiovisual trials (23ms). Unlike TAC, and

contrary to our predictions, the MSE effect was not further modulated by either contextual

factor, with no 2-way interaction with MR, $F < 1$, and the 2-way interaction with DO showing

only a statistically nonsignificant trend, $F_{(1, 38)}$ = 3.6, p = 0.07, η_p^2 = 0.1.

Figure 3. Overall contra- and ipsilateral ERP waveforms representing a mean amplitude over electrode clusters

(plotted on the head model at the bottom of the figure in blue and black [in light blue and dark grey,

respectively, ms-by-ms error areas representing standard error of the mean, SEM]), separately for each of the

four experimental conditions (Distractor Colour x Distractor Modality), averaged across all four adult

experiments. The N2pc time-window of 180–300ms following distractor onset is highlighted in grey, and

significant contra-ipsi differences are marked with an asterisk ($p < 0.05$). As expected, only the TCC distractors
elicited statistically significant contra-ipsi differences.

Unexpectedly, the overall extent of multisensory enhancement depended on the
colour of the distractor, evidenced by a 2-way Distractor Colour × Distractor Modality
interaction, $F_{(1, 38)} = 6.4$, $p = 0.015$, $\eta_p^2 = 0.2$. This effect was driven by behavioural capture
effects elicited by TCC distractors to be larger on AV (38ms) than V trials (46ms), $t_{(38)} = 4.3$, p
< 0.001; there was no evidence for similar effects for NCC distractors, $t_{(38)} = .3$ $p = 0.78$; see
Supplemental Figure 1). However, this pattern was not observed within each experiment
individually (all F's < 3, all p's > 0.1; see SOMs for full details). Behavioural capture effects
were also jointly modulated by the two contextual factors independently of TAC and MSE,
$F_{(1, 38)} = 8.3$, $p = 0.007$, $\eta_p^2 = 0.2$. As these interactions were outside of our main research
questions focusing on TAC- and MSE-bound interactions with the contextual factors, we
describe them in full in the SOMs.

ERP analyses

The role of lateralised (N2pc-like) brain mechanisms

The role of lateralised, N2pc-like brain mechanisms in the interactions between TAC and
MSE on the one hand and the two studied contextual factors on the other hand was limited,
even with the added use of EN analyses. The full discussion of these results can be found in
the SOMs, while we only briefly summarise them here. Canonical N2pc analyses on the
mean amplitudes confirmed TAC, with larger mean N2pc for TCCs (-0.08µV) than for NCCs (-
0.02µV; plotted for V and AV conditions for reference, Figure 3). Both GFP and topographic
analyses of the lateralised ERPs also showed evidence of TAC. However, there was no
evidence for the modulation of TAC by contextual factors (in N2pc, F's < 3, p's > 0.1, in GFP
and topography, all F's < 1). For MSE, behavioural modulations were not mirrored in either
canonical N2pc analyses, GFP or topographic analyses; similarly, there was no evidence for
modulation of MSE by contextual factors in any of these measures. Neither the main effects
of MR and DO, nor the 2-way MR x DO interaction were significant in the lateralised
analyses of N2pc, GFP, or ERP topography (all F's < 1). Map presence in the N2pc time-
window was modulated jointly by the two contextual factors; we describe those results in
SOMs.

The role of nonlateralised brain mechanisms

The major part of our analyses focused on investigating the role of nonlateralised ERP

mechanisms in contributing to the interactions between the multisensory, stimulus-driven

and visual, goal-based attentional control and contextual control. To remind the reader, to

prevent these nonlateralised ERPs from being confounded by the presence of sound on AV

trials, we based the current analyses on difference ERPs indexing visual attentional control

under sound presence vs. absence. Specifically, these ERPs reflected the difference between

TCCAV and NCCAV conditions, and between TCCV and NCCV conditions (D_{AV} and D_V levels,

respectively, of the factor Target Difference). We focus the description of these results on

the effects of interest (see SOMs for full description of results).

The 2 × 2 × 2 rmANOVA on electrode-wise voltage analyses revealed that the main

effect of Target Difference was significant at 53–99ms and 141–179ms, where amplitudes

were larger D_{AV} (TCCAV − NCCAV difference) than for D_V (TCCV − NCCV difference). This

effect was further modulated by the multisensory relationship within the distractors, with a

2-way Target Difference × MR interaction, at the following time-windows: 65–103ms, 143–

171ms, and 194–221ms (all p's < 0.05). This effect was driven by semantically congruent

distractors showing larger amplitudes for D_{AV} than D_V within all 3 time-windows (65–97ms,

143–171ms, and 194–221ms; all p's < 0.05). No similar differences were found for

simultaneous distractors, and there were no other interactions.

Interaction of TAC with contextual factors. We next used EN analyses to investigate the

contribution of the strength- and topography-based nonlateralised mechanisms to the

interactions between TAC and contextual factors.

Strength-based brain mechanisms. A 2 × 2 × 2 rmANOVA on the GFP, mirroring the

electrode-wise analysis on ERP voltages, also showed the main effect of Target Difference

spanning a large part of the first 300ms post-distractor (19–213ms, 221–255ms, and 275–

290ms), where, just like the voltages, also GFP was larger for D_{AV} than D_V (all p's < 0.05).

In GFP, Target Difference interacted both with MR (23–255ms) and separately with

DO (88–127ms; see SOMs for full description), but most notable was the 3-way Target

Difference × MR × DO interaction, spanning 102–124ms and 234–249ms. We then followed

up this interaction with series of post-hocs to gauge the modulations of TAC (and MSE, see

below) by the two contextual factors. In GFP, MR and DO interacted independently of

Target Difference in the second time-window, which results we describe in the SOMs.

To gauge differences in the strength of TAC in GFP across the 4 contexts, we then

compared directly only the visually-elicited target differences across respective levels of the

2 contextual factors. Weakest GFPs were observed for simultaneous predictable distractors

(Figure 4A). They were weaker than GFPs elicited by simultaneous unpredictable distractors

(102–124ms and 234–249ms), and predictable congruent distractors (only in the later

window, 234–249ms).

Nonlateralised mechanisms

TAC-related mechanisms

A) GFP

Simultaneous Unpredictable
Simultaneous Predictable
Congruent Unpredictable
Congruent Predictable

B) ERP Topography

Map A1   Map A2   Map A3   Map A4   Map A5

29–126ms: 3-way Map x MR x DO interaction

Figure 4. Nonlateralised GFP and topography results for the visual only difference ERPs (DV condition of Target

Difference), as a proxy for TAC. **A)** Mean GFP over the post-distractor and pre-target time-period across the 4

experimental tasks (as a function of the levels of MR and DO that they represent), as denoted by the colours

on the legend. The time-windows of interest (102–124ms and 234–249ms) are highlighted by grey areas. **B)**

Template maps over the post-distractor time-period as revealed by the segmentation (Maps A1 to A5) are

shown in top panels. In lower panels are the results of the fitting procedure over the 29–126ms time-window.

The results displayed here are the follow-up tests of the 3-way Map x MR x DO interaction as a function of MR

(leftward panel) and of DO (rightward panel). Bars are coloured according to the template maps that they

represent. Conditions are represented by full colour or patterns per the legend. Error bars represent SEMs.

Topography-based brain mechanisms. We focused the segmentation of the TAC-

related topographic activity on the 0–300ms post-distractor time-window (before the target

onset), which revealed 10 clusters that explained 82% of the global explained variance

within the visual-only ERPs. This time-window was largely composed of distinct template

maps across the 4 context conditions, except a time-window of 29–126ms post-distractor

where segmentation revealed template maps shared across conditions. A 5 × 2 × 2

rmANOVA on the map presence over the 29–126ms post-distractor time-window revealed a

3-way Map x MR × DO interaction, $F_{(3.2, 122)} = 5.3$, $p = 0.002$, $\eta_p^2 = 0.1$.

Follow-up tests focused on maps differentiating between the 4 conditions, first as a

function of MR. Simultaneous predictable distractors, which elicited the weakest GFP,

activated one map (map A2), and did so for reliably shorter time (18ms) compared to

simultaneous unpredictable (35ms) distractors, $t_{(38)} = 2.96$, $p < .01$. The same map showed

the opposite pattern for semantically congruent distractors, where it instead predominated

responses to predictable (25ms) vs. unpredictable (14ms) distractors, $t_{(38)} = 2$, $p = 0.04$

(Figure 4B, leftward panel). In contrast with this pattern, a second map (map A5)

predominated responses to semantically congruent unpredictable (34ms) vs. semantically

congruent predictable (19ms) distractors, $t_{(38)} = 2.7$, $p = 0.04$ (Figure 4B, leftward panel).

Follow-ups conducted as a function of DO revealed that the simultaneous

predictable distractors activated a third map (map A1), which predominated responses to

these distractors (37ms) vs. semantically congruent predictable distractors (21ms), $t_{(38)} = 2.7$,

$p < 0.01$ (Figure 4B, rightward panel). Map A2 also predominated responses to

unpredictable simultaneous distractors (35ms) vs. unpredictable semantically congruent

(14ms) distractors, $t_{(38)} = 3.7$, $p = 0.001$ (Figure 4B, rightward panel). And Map A5 again

predominated responses to unpredictable semantically congruent (34ms), now compared to

unpredictable simultaneous (12ms) distractors, $t_{(38)} = 3.7$, $p < 0.001$.

Figure 5. Nonlateralised GFP and topography results for the difference ERPs between the D_{AV} and D_V conditions of Target Difference, as a proxy for MSE. **A)** Mean GFP over the post-distractor and pre-target time-period across the 4 experimental tasks (as a function of the levels of MR and DO that they represent), as denoted by the colours on the legend. The time-windows of interest (102–124ms and 234–249ms) are highlighted by grey bars. **B)** Template maps over the post-distractor time-period as revealed by the segmentation (Maps A1 to A7) are shown on top. Below are the results of the fitting procedure over the three

time-windows: 35–110, 110–190, and 190–300 time-window. Here we display the follow-ups of the

interactions observed in each time-window: in 35–110 and 190–300 time-windows, the 2-wau Map x MR

interaction (leftward and rightward panels, respectively), and in the 110–190 time-window, follow-ups of the

3-way Map x MR x DO interaction as a function of MR and of DO (middle panel). Bars are coloured according

to the template maps that they represent. Conditions are represented by full colour or patterns per the

legend. Error bars represent SEMs.

Interaction of MSE with contextual factors. We then analysed the strength- and topography-

based nonlateralised mechanisms contributing to the interactions between MSE and

contextual factors.

Strength-based brain mechanisms. To gauge the AV-induced MS enhancements

between D_{AV} and D_V across the 4 contexts, we explored the abovementioned 2 × 2 × 2 GFP

interaction using a series of simple follow-up post-hocs. We first tested if the response

strength between D_{AV} and D_V was reliably different within each of the 4 contextual

conditions. We found that enhanced AV-induced target ERP responses characterised

primarily semantic distractors. That is, for both predictable and unpredictable semantically

congruent distractors, GFP were larger for D_{AV} than D_V across both earlier and later time-

windows. However, for simultaneous distractors, D_{AV} and D_V were larger only in predictable

contexts and only in the earlier (102–124ms) time-window; no differences were found at all

for unpredictable distractors.

Next, to gauge the AV-induced MS enhancements between D_{AV} and D_V across the 4

contexts, we calculated the subtraction between D_{AV} and D_V for each context, and

compared these across the respective levels of the 2 factors. AV-induced enhancements

were weaker for predictable simultaneous distractors than predictable semantically

congruent distractors (102–124ms and 234–249ms;

Figure 5A). The AV-enhancements for predictable simultaneous were weaker also

compared to unpredictable simultaneous distractors in both time-windows but these did

not meet the temporal criteria for statistical significance.

Topography-based brain mechanisms. We focused the segmentation of the MSE-

related topographic activity in the 0–300ms post-distractor time-window, which revealed 7

clusters that explained 78% of the GEV within the AV-V target difference ERPs. We fit the

data in three subsequent time-windows where there were clear topographic patterns, 35–

110ms, 110– 190ms, and 190– 300ms. To foreshadow the results, in the first and third time-

window the MSE-related templates were modulated only by MR, while in the middle time-

window – by both contextual factors.

In the first, 35–110ms time-window, the modulation of map presence by MR was

evidenced by a 2-way Map × MR interaction, $F_{(2.1,77.9)}$ = 9.2, p < 0.001, η_p^2 = 0.2. This effect

was driven by one map (map B3) that in this time-window predominated responses to

semantically congruent (42ms) vs. simultaneous (25ms) distractors, $t_{(38)}$ = 4.3, p = 0.02,

whereas another map (map B5) predominated responses to simultaneous (33ms) vs.

semantically congruent (18ms) distractors, $t_{(38)}$ = 4, p = 0.01 (

Figure 5B leftward panel).

In the second, 110– 190ms time-window, map presence was modulated by both

contextual factors, with a 3-way Map × MR × DO interaction, $F_{(2.6,99.9)}$ = 3.7, p = 0.02, η_p^2 =

0.1.

Map B6 predominated responses to *semantically congruent unpredictable* (37ms) vs.
semantically congruent predictable distractors (21ms, $t_{(38)}$ = 2.5, p = 0.02;
Figure **5**B middle top panel), and vs. simultaneous unpredictable distractors (24ms, $t_{(38)}$ =
2.3, p = 0.04;
Figure **5**B bottom middle panel). Map B5 predominated responses to *simultaneous*
unpredictable (33ms) vs. simultaneous predictable distractors (17ms, $t_{(38)}$ = 2.6, p = 0.04;
Figure **5**B middle upper panel), and vs. semantically congruent unpredictable distractors
(13ms, $t_{(38)}$ = 3.4, p = 0.002;
Figure **5**B middle lower panel). Map B3 predominated responses to *semantically congruent*
predictable (25ms) vs. semantically congruent unpredictable distractors (12ms, $t_{(38)}$ = 2.2, p =
0.05;
Figure **5**B middle top panel), and vs. predictable simultaneous distractors (8ms, $t_{(38)}$ = 2.2, p =
0.005;
Figure **5**B middle lower panel). Finally, map B1 predominated responses to *simultaneous*
predictable (35ms;
Figure **5**B middle upper panel) vs. simultaneous unpredictable distractors (18ms, $t_{(38)}$ = 2.8, p
= 0.01), and vs. semantically congruent predictable distractors (17ms, $t_{(38)}$ = 2.8, p = 0.01;
Figure **5**B middle lower panel).

In the third, 190–300ms time-window, the 2-way Map × MR interaction was reliable

at $F_{(3.2,121.6)}$ = 3.7, p = 0.01, η_p^2= 0.1. Notably, the same map as before (map B3)

predominated responses to semantically congruent (50ms) vs. simultaneous distractors

(33ms), $t_{(38)}$ = 3.6, p = 0.08, and another map (map B1) predominated responses to

simultaneous (25ms) distractors vs. semantically congruent (14ms) distractors, $t_{(38)}$ = 2.3, p =

0.02 (

Figure 5B rightward panel).

838   **Discussion**

Real-world environments are cluttered with competing information and necessitate

effective attentional control in order to focus on relevant inputs and ignore unimportant

inputs. In such environments, goal-based top-down and bottom-up processes interact with

context-based processes of attentional control. Yet these types of control processes have

traditionally been studied in isolation. Here, for the first time, we studied these control

processes together, as they interact in multisensory settings. First, we studied how visual-

top down control (as indexed by task-set contingent attentional capture, or TAC) interacts

with contextual factors, like semantic relationships within distractor features and their

temporal predictability. Second, we investigated how such interactions transpire for

multisensory bottom-up attentional control (as indexed by multisensory enhancement of

visual attentional capture, or MSE). Finally, we discuss our findings in the context of

theoretical proposals regarding how contextual factors interact with visual attentional

control, suggesting how these the interactions may occur in more naturalistic, real-world

settings.

Interaction of task-set contingent attentional capture with contextual control

We revealed several new findings regarding how visual attentional control operates in

settings that are rich in contextual and multisensory information. Most importantly, we

found that visual top-down goal-based attentional control indexed by task-set contingent

attentional capture (TAC) interacted with both contextual factors. At the level of behaviour,

these interactions involved multisensory relationship (MR) and transpired by target-colour

distractors capturing attention more strongly when they were simultaneous rather than

semantically congruent. In turn, multisensory relationship did not modulate behavioural

capture by nontarget-colour distractors. This pattern of results suggests that the brain

utilises context-related information for goal-directed behaviour also in multisensory

contexts.

The EN analyses of the nonlateralised ERPs across the whole post-distractor time-
window both confirmed and enriched our behavioural findings. Notably, this contrasted
with the canonical lateralised, N2pc-like brain mechanisms, which revealed no modulations
of TAC by either contextual factor. First, in terms of findings in lateralised brain mechanisms,
target difference ERPs from visual-only conditions (reflecting the TAC) showed especially
attenuated responses (indexed by lower GFP) to simultaneous distractors that were
predictable. They were reliably weaker than simultaneous *unpredictable* distractors and
semantically congruent predictable distractors. Second, it is notable that these TAC–context
interactions occurred both in the time-window of the N2pc, i.e., during the 'attentional
selection of candidate target objects' (Eimer, 1996, 2014), but also earlier. Namely, the
suppression of GFP to simultaneous predictable vs. simultaneous unpredictable distractors
were observed already in the early, 102–124ms (and later, N2pc-like, 234–249ms time-
window) time-window. Third, segmentation analyses showed that simultaneous predictable
distractors activated two distinct configurations of brain sources (reflected by two distinct
templates) already ~30ms post-distractor. One map (map A1) predominated responses to
these distractors, compared to predictable semantically congruent distractors; notably, the
second map (map A2) activated by simultaneous predictable distractors was involved in
their processing to reliably lesser extent than that of simultaneous *unpredictable*
distractors.

To summarise, target-colour visual distractors capture attention more strongly when
they are simultaneous than semantically congruent. This effect co-occurred with a range of
findings in nonlateralised brain mechanisms spanning both the attentional selection, N2pc-
like time-window as well as much earlier time-windows. Simultaneous distractors, but only
predictable ones, activated one network in the early time-window for longer period of time,
and another network for reliably shorter period of time, compared to semantically
congruent distractors. These network-based processes were followed by strength-related
processes, where the same, simultaneous (again, predictable) distractors elicited the most
attenuated responses (compared to both semantically congruent predictable and
simultaneous unpredictable distractors). This cascade of modulations over nonlateralised
brain mechanisms might offer a potential mechanisms whereby goal-directed attentional
control utilises contextual information. While in our current paradigm, behaviourally these
effects were more prominent for the semantic relationships within the distractors, our

multivariate EN analyses of the nonlateralised mechanisms revealed the importance of both

semantic and temporal prediction-based processes, which transpire across both strength-

based and network-based mechanisms, and from early on.

To our knowledge, present findings are the first demonstration of how goal-driven

feature-based attentional control interacts with multiple sources of contextual control. They

both confirm and extend existing understanding of how predictions affect perception and

action, as well as attentional control. For one, the early (~30ms) suppressed responses to

predictable target-colour (compared to unpredictable) stimuli are in line with the early-

onset (~50ms post-stimulus) suppressions of visual ERPs to temporally predictable stimuli

(C1; Dassanayake et al. 2016). Our EN analyses revealed that the earliest suppression at

least in some contexts might be driven by weaker activation of relevant brain networks,

which at later stage (~100ms post-stimulus) also involve suppressed brain responses. The

latency of those later, strength-based suppressions are in line with attenuations of N1

activity for self-generated sounds (Baess et al. 2011; Klaffehn et al. 2019), and around the

time when the brain promotes the processing of unexpected events according to recent

theories (e.g., Press et al. 2020). Notably, the latency of our later, strength-based effects is

also in line with the earliest latency (~115ms post-stimulus) of the goal-based suppression of

salient (i.e., presented at the same time as targets) visual distractors (distractor positivity,

Pd; Sawaki & Luck 2010). Van Moorselaar and Slagter (2019) showed that when such salient

visual distractors appear in predictable locations, they elicit N2pc but no longer a Pd,

suggesting that learning the location of the distractor enables the brain to effectively ignore

it, without the need for active inhibition (see Matusz et al. 2016 for similar findings showing

spatial predictability inhibiting attention shifts to irrelevant sounds). More recently, van

Moorselaar et al. (2020b) showed that two types of predictions regarding the distractor,

such as that of its location and of its feature, interact, with feature-based prediction

attenuating location-based suppression. Notably, representation of the predictable

distractor feature could be decoded already from pre-stimulus activity. While the two

paradigms differ from each other, our results might potentially enrich those findings. Our

task is not optimised for measuring Pd (as distractors appear at irrelevant moments in time),

but the target-colour distractors therein reliably capture attention (Folk et al. 1992). As we

have shown, semantic relationships might be a mechanism through which this capture can

be at least partly suppressed. Thus, our findings suggest that, attentional control will flexibly

utilise all available contextual information - different types of predictions (spatial, feature-
based, temporal, etc.; e. van Moorselaar & Slagter 2020b) and semantic relationships - to
facilitate goal-directed processing.

While this was not the main aim of the present study, it is notable that we observed
the task-set contingent attentional capture despite the addition to the paradigm of both the
multisensory manipulation (as in Matusz & Eimer 2011) and the two contextual factors. As
in traditional literature, target-colour distractors triggered reliable behavioural capture
effects and enhanced mean N2pc amplitudes compared to nontarget-colour distractors
(Eimer et al., 2009, Eimer & Kiss, 2008, Kiss et al., 2008a; 2008b). This suggests that this
form of visual top-down attentional control of object attentional selection, based on target
features (like colour), is an effective, robust form of attentional control, interacting with
both multisensory and contextual mechanisms.

Interaction of multisensory enhancement of attentional capture with contextual control
Our results offer an important replication of the multisensory enhancement of attentional
capture (MSE) effect demonstrated in a rigorous paradigm by Matusz and Eimer (2011). In
the current setup and unlike the original study (Matusz & Eimer 2011), multisensory
enhancement of attentional capture did interact with colour cue. Given the strong feature-
specific visual top-down attentional demands of the task, the evidence of behavioural
multisensory enhancement supports the idea put forward by Matusz and Eimer (2011) that
multisensory enhancements of visual attentional capture can occur involuntarily and
independently of task demands (see also Matusz et al. 2015). We did indeed observe again
the two-way interaction that let the authors make such a conclusion. In contrast, here, this
effect was indeed dependent on the colour matching between target and the cue. Namely,
the MSE of attentional behavioural capture was found for target-colour distractors, but not
for the nontarget-colour distractors. However, when we tested for this 3-way interaction
within each of the experiments, this interaction was no longer present. It was not present
even in Experiment 2 that was the closest to the original study, as in both studies distractors
were temporally predictable and simultaneous. While care is needed when interpreting
these null results, one possibility is that in fact in the real world bottom-up multisensory and
top-down goal-directed attentional control processes may interact - where goal-based
attentional control processes would also utilise the increased salience of simultaneous

multisensory objects. However, it is possible that such interactions are subtle and/or the
current paradigm is not optimal for gauging them.

While we again found multisensory enhancement of visual attentional capture
(which interacted with TAC), its interactions with the two contextual factors did not
transpire at the level of behaviour. Notably, the lateralised ERPs that are traditionally
analysed in attentional control research and the canonical N2pc analyses that gauge them
appear not to be sensitive to multisensory enhancements altogether. To date, there is weak
evidence for distractor-based bottom-up effects of multisensory enhancement in lateralised
ERPs (van der Burg et al. 2011). Lateralised ERPs (and N2pc in particular) more readily reveal
influences of top-down goal-based multisensory processes (Matusz & Eimer 2013; Matusz et
al. 2019). The contribution of lateralised brain mechanisms to the interactions between
multisensory enhancements of visual attention capture and contextual processes might be
equally limited. As in the case of top-down visual attentional control, nonlateralised brain
processes contribute to those interactions more prominently. We discuss those findings
below in detail.

First, similarly to visual attentional control, nonlateralised brain mechanisms were
involved in multisensory processing from early on after distractor onset and included both
strength-based and network-based (nonlateralised) brain mechanisms. Specifically, the
Target Difference factor that was part of the 3-way interaction with the two contextual
factors, contained both visual and multisensory target-difference (TCC and NCC difference).
When we analysed the AV–V difference ERPs as a proxy to measure processing related to
multisensory enhancement, we identified reliably stronger (indexed by GFP) multisensory
than visual ERPs for simultaneous predictable (but not unpredictable) distractors (in the
earlier 102-124ms time-window), and for both predictable and unpredictable semantically
congruence distractors, across both earlier and later (234-249ms) time-windows. Notably, in
both time-windows audiovisually-induced enhancement of ERPs were reliably larger for
predictable and semantically congruent than predictable and simultaneous distractors. This
pattern of results provides a novel insight into attentional selection of audio-visual objects.
Specifically, even when such objects are irrelevant to the task, multisensory relationships
modulate their processing, they do so from earlier, perceptual to later, cognitive stages, and
through interactions with other contextual factors, such as stimulus (temporal)
predictability.

Second, network-related brain mechanisms contributed to the interactions between
multisensory enhancements and the two contextual factors and did so throughout the post-
distractor time-period. Semantic relationship played a predominant (but not absolute) role
in those nonlateralised response interactions. The AV–V difference ERPs in the first, early
(35–110ms time-window) and late, N2pc-like (190–300ms) time-windows were both
modulated exclusively by multisensory relationships. Therein, the same map predominated
the responses to semantically congruent as compared to simultaneous distractors, while a
different map in the two time-windows was predominated those to simultaneous
distractors. This result suggests that both early perceptual processing and later, attentional
selection processing is modulated by meaning (vs lack thereof). The fact that this was the
same template map (map B3) suggests that a single network of brain sources was activated
for a longer time period at both stages to promote the processing of meaningful
multisensory stimuli. In contrast, in the second, 110–190ms time-window, the AV–V
difference ERPs was modulated by both multisensory relationship and distractor
predictability. The same "semantic" template map was present also in this mid-latency time-
window, now predominating the response for semantically congruent predictable versus
unpredictable distractors. Notably, the two maps that were present in the mid-latency time-
window for simultaneous distractors and modulated further as a function of DO, were also
present in the early time-window (map B5) and in the late time-window (map B1). Together,
these findings suggest that from early on distinct types of networks are activated for
meaningful and meaningless stimuli, even when those stimuli are not relevant to the
current task and goals. We know that semantic relationships can improve object recognition
(e.g. animals) even when only one of the senses is task-relevant (Yuval-Greenberg &
Deouell, 2007). Similar benefits are found in unisensory (visual, auditory) in a continuous
old/new recognition task (Murray et al. 2004; Matusz et al. 2015a; Thelen et al. 2015),
where benefits from mere multisensory simultaneity shows strong inter-individual
variability (Thelen et al. 2014). The current results enrich those findings by showing that in
naturalistic settings, object processing (attentional selection) will be jointly affected by
bottom-up multisensory processes and one's goals, but also contextual factors, such as
stimulus predictability. Furthermore, these influences likely transpire across both task-
relevant and task-irrelevant objects, contrary to what traditional research focused
predominantly on target processing might suggest.

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How we pay attention in naturalistic settings

It is well-recognised that the brain can process and act on a very limited portion of the

incoming stimuli. One mechanism to aid resolution of this challenge is goal-based

attentional control that promotes the processing of stimuli matching our currently set

behavioural goals, based on stimulus features or their location in space or time (Folk et al.

1992; Gazzeley & Nobre 2012; Noonan et al. 2018). It is now well-established that the brain

likewise utilises regularities in the outside world to help itself reduce the complexity of the

external stimulation, by creating expectations that constrain the potential interpretation of

the incoming stimuli (Summerfield & Egner 2009; Press et al. 2020). It is likewise

increasingly recognised that the semantic relationships between objects and events in

everyday situations stored in the long-terms semantic memory impact stimulus processing

and attentional selection (Summerfield et al. 2006; Peelen & Kastner 2014). While these

theoretical accounts have been invaluable in clarifying the mechanisms that support goal-

directed behaviour in everyday situations, these mechanisms (and their interactions with

goal-based attentional control) have been studied systematically in isolation in the visual

(Summerfield & Egner 2009; van Moorselaar & Slagter 2019; van Moorselaar et al. 2020b),

or the auditory (e.g., Chennu et al. 2013) domain. Here, we showed that more traditionally

studied attentional control mechanisms, such as feature-based goal-driven attention control

(indexed by TAC) and salience-based (here, multisensory) attention control mechanisms

interact with contextual mechanisms including both stimulus predictability and semantic

relationships.

In line with existing visual literature, visual goal-based attentional control interacted,

behaviourally, with semantic relationships induced in the audio-visual stimuli half-way

through the experiment. Namely, distractors that matched the colour of the target and so

the goals captured attention *less* when they contained meaning, compared to when only

simultaneity joined visual and auditory distractor stimuli. This is an important finding as it

suggests that goal-based control by stimulus feature (like colour) may, indeed, involuntarily

extend to even task-irrelevant stimuli, as postulated originally by Folk and colleagues (1992)

and replicated numerous times. Here, we revealed that equipping task-irrelevant stimuli

with meaning may provide the brain with an otherwise unavailable mechanism to improve

goal-based attentional control. In our task, distractors appear at task-irrelevant moments in

time and uninformative about subsequent spatial location of the target, and so it is
beneficial to inhibit their processing if possible. Thus, more generally, in everyday situations,
where many objects likely match features with the objects we are currently searching, their
ability to capture our attention might be *de facto* weaker than what the traditional
literature on visual attentional control suggests.

Our EEG results demonstrated that semantically congruent and simultaneous stimuli
preferentially activate distinct brain networks as early as 30ms after onset of the distractors.
Therein, the temporal predictability further differentiated the brain networks activated by
semantically congruent and simultaneous distractors; however, modulations stemming from
predictions were not reliably visible in our behavioural measures of attentional capture. It is
important to note that context-based interactions of meaning and predictability influenced
the brain networks activated by target-congruent and nontarget-congruent visual
distractors as early as ~30ms – before any attentional selection-related activity is expected
to happen. While our analyses prevent strong conclusions, it is possible that these
differently activated early brain networks led to the stronger brain responses at both 100ms
and 250ms post-distractor, the latter effects matching (later parts of) the time-window of
spatially-selective responses indexed by the N2pc. These findings suggest that, in everyday
environments, attentional control will utilise all available contextual information about
visual stimuli and their context (semantic relationships; predictions) in the service of goal-
related processing. These influences will occur already during early stages of brain activity,
with at least some of those modulations transpiring also at the level of behaviour. In other
words, meaning may be an important - yet strongly under-researched - type of contextual
information utilised by goal-based attentional control in both unisensory (Chennu et al.
2013; Peelen & Kastner 2014; Retsa et al. 2020) and multisensory (Matusz et al. 2016;
Sarmiento et al. 2016; Retsa et al. 2018) settings. Such findings provide important new
insights into how attentional control operates in real-world environments, where space-
and feature-based attentional control interact routinely with contextual processes. Notably,
all these influences that we have observed were measured on task-*irrelevant* stimuli,
suggesting an even more complex interplay between goal-based attentional control and
distraction than one proposed by Folk and colleagues (1992), but one where both task-
based and contextual information is utilised simultaneously to control our attention to both

task-relevant and task-irrelevant objects (ten Oever et al. 2016; Noonan et al. 2018; van

Moorselaar & Slagter 2020a).

Our nonlateralised global EEG analyses revealed robust interactions between

bottom-up multisensory attentional control and contextual influences. Distinct brain

networks were involved predominantly in attention towards semantically congruent and

simultaneous multisensory stimuli. These modulations began early on during distractor

processing and transpired throughout the time-window of N2pc-like attentional processes.

While these interactions were not visible at the level of behaviour, they are nevertheless

notable. This is because these interactions occurred despite the paradigm involving

stringent, visual feature-specific attentional control settings and thus was more optimised

for abolishing multisensory enhancement of visual attentional capture. Not only was

multisensory enhancement not eliminated, it was replicated (Matusz & Eimer 2011) and

shown to interact with both stimulus meaning and stimulus predictability. This pattern of

results suggest that it is a robust finding and thus needs to be considered when aiming to

study attentional control and information processing more generally in naturalistic settings

(e.g. such as during viewing movies or listening to audiostories) or veridical real-world

environments (e.g. the classroom or the museum). Our results suggest that in such contexts

person's goals will interact with the contextual information (stimulus meaning, different

forms of prediction) but also with dynamic changes in stimulus perceptual salience driven by

moment-to-moment bottom-up multisensory enhancement. The strength of these

modulations in everyday situations can be at least partly gauged through naturalistic studies

in the laboratory (van Atteveldt et al. 2018; Matusz et al. 2019a), such as the present one,

where task-relevance, attention demands and context influences are systematically varied

together with manipulations of the engaged senses. In more naturalistic settings, the same

effects could be measured, and thus controlled, by systematically comparing, where

possible, unisensory and multisensory stimulus or event presentations. Such a cyclical

(Matusz et al. 2019a) approach involving systematic testing of hypotheses across laboratory

and outside settings is most likely to bring us closer to successfully bridging the laboratory-

developed (typically visually-derived) hypotheses and theories of attentional control and

other cognitive functions and the recently accumulating findings from such naturalistic

setups. More generally, our findings suggest that conceptualisations of ecological validity

and naturalistic settings (Peelen et al. 2014; Shamay-Tsoory & Mendelsohn 2019; Vanderwal

et al. 2019; Eickhoff et al. 2020; Cantlon 2020) could benefit from addition to more

traditionally invoked components (e.g. attention demands, context, socialness) also that of

multisensory influences.

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<p style="text-align:center">**References**</p>

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Author contributions

Nora Turoman: Investigation, Formal analysis, Data curation, Software, Visualisation, Writing - original draft.

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Micah M. Murray: Funding acquisition, Methodology, Resources, Formal analysis, Software, Supervision, Writing - review & editing.

Conflict of interest statement

1472   The authors declare that the research was conducted in the absence of any commercial or

1473   financial relationships that could be construed as a potential conflict of interest.

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