Title:

Can expectation suppression be explained by reduced attention to predictable

stimuli?

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Abstract

The expectation-suppression effect – reduced stimulus-evoked responses to expected

stimuli – is widely considered to be an empirical hallmark of reduced prediction errors

in the framework of predictive coding. Here we challenge this notion by proposing that

this phenomenon can also be explained by a reduced attention effect. Specifically, we

argue that reduced responses to predictable stimuli can also be explained by a reduced

saliency-driven allocation of attention. To resolve whether expectation suppression is

best explained by attention or predictive coding, additional research is needed to

determine whether attention effects precede the encoding of expectation violations (or

vice versa) and to reveal how expectations change neural representations of stimulus

features.

Introduction

Over the past decade, predictive coding theories have gained significant traction with

high-profile findings across a range of cognitive domains (Friston, 2005; Rao & Ballard,

1999; Summerfield & de Lange, 2014; Walsh et al., 2020). Central to this theory lies

the idea that the computation of prediction errors forms a principal computational

mechanism in the brain. One phenomenon that has been frequently reported and cited

as empirical support for such theories is the expectation-suppression phenomenon –

reduced stimulus-evoked responses to expected stimuli (Alink et al., 2010; Blank &

Davis, 2016; Richter et al., 2018; Todorovic & de Lange, 2012; Walsh & McGovern,

2018). The exact neural mechanism via which predictive coding could give rise to

expectation suppression, however, remains unclear. For example, a debate is still

ongoing on whether it results from dampening (or explaining away) of neural

representations of predictable sensory input or from more selective (or 'sharpened')

neural representation of predictable stimuli (Friston, 2005; Kok et al., 2012; Richter et

al., 2018; Walsh & McGovern, 2018). In this perspective, we want to discuss the

alternative interpretation that expectation suppression results from reduced attention

to predictable stimuli (Aitchison & Lengyel, 2017). This would be in line with the view

that surprising stimuli are more salient and more likely to attract attention (Desimone

& Duncan, 1995; Itti & Baldi, 2009).

Two alternative explanations for expectation suppression

A prominent view suggests that expectation suppression occurs irrespectively of whether stimuli are attended (den Ouden et al., 2009; Kok et al., 2012; Summerfield & Egner, 2009, 2016). One argument for this notion was that better detection of predictable stimuli paired with reduced stimulus evoked responses is difficult to reconcile with the positive neural gain mechanism associated with attention (Alink et al., 2010; Treue & Trujillo, 1999).

This independency between attention and expectation has been challenged by a recent study by Richter and de Lange (2019). They showed that diverting attention away from a predictable image completely abolishes the expectation-suppression effect in the visual cortex. Specifically, conditional probabilities of presented object images no longer had an effect on the amplitude of visual cortex responses if the participants' attention was drawn away from the object images by a central fixation task. Nonetheless, this finding is interpreted in light of a ***predictive-coding based explanation*** (PBE) by proposing that attending an object boosts prediction error coding. This interpretation is in line with the predictive coding framework, in which attention increases the gain of prediction error units (Feldman & Friston, 2010; Friston, 2009; Spratling, 2008). In addition, Richter and de Lange suggest that the generation of predictions depends on attention. Note that, in contrast to this attention-driven boost of error coding, attention has also been reported to suppress prediction error coding (Sohoglu & Chait, 2016).

In this perspective, we want to discuss an alternative ***attention-based explanation*** (ABE) of expectation suppression which attributes reduced responses to predictable stimuli to a reduced allocation of attentional resources. This alternative is motivated by the notion that attention results from a competition between stimulus representations for the limited information processing capacity of the brain (Desimone & Duncan, 1995). This competition is biased towards behaviourally relevant stimuli, e.g., via enhanced responses to task-relevant or salient stimuli. According to this view, saliency is a stimulus characteristic that automatically results from deviations from both its physical context and from 'higher-level' context based on natural image statistics and experience (Desimone & Duncan, 1995, Kanan et al., 2009). Importantly, the saliency of unpredictable stimuli depends on context: a stimulus is more salient when it is the deviant stimulus in an oddball paradigm as compared to when it is part of a

random stimulus sequence because only the former would result into a contextual
violation. Accordingly, ABE proposes that enhanced responses to unexpected stimuli
in sensory brain areas reflect a greater saliency-driven allocation of attentional
resources towards these stimuli (Figure 1). Note that, in contrast to the PBE, the ABE
does not strictly require prediction-error coding as the expectation suppression effect
is fully explained by reduced stimulus saliency (but see Spratling, 2008). Finally, we
would like to point out that the ABE does not preclude the possibility that attention can
be drawn towards predictable events under certain circumstances (e.g., when
predictable stimuli are task-relevant, Alamia & Zénon, 2016).

Figure 1 Predictive-coding based explanation (PBE, left) attention increases the responsiveness of neurons sensitive to the discrepancy of the sensory input and the expected stimulus. **Attention-based explanation** (ABE, right) explains expectation suppression as resulting from a reduced stimulus saliency.

For illustrative purposes, let us evaluate whether the ABE can accommodate
the finding by Richter and de Lange (2019) that expectation suppression only occurs
in the visual cortex when images are attended. As outlined above, unexpected stimuli
can be thought of as more capable of capturing attention due to their saliency.
Therefore, predictable images should have received less attention which could explain
the reduced associated responses in the visual cortex. The reported absence of this
expectation-suppression effect for the central fixation task also fits with the ABE as this
task is argued to completely prevent attention allocation towards expected and
unexpected images (Richter & de Lange, 2019).

Can attention and expectation effects be orthogonalized?

An influential view is that it is possible to orthogonalize feature-based attention and
expectation effects in experimental designs (Summerfield & Egner, 2016). However,
as pointed out by Summerfield and Egner (2009, 2016), many studies on predictive

coding still contain subtle dependencies between them (see also
Rungratsameetaweemana & Serences, 2019). For example, the experimental design
used by Richter and de Lange (2019) allowed participants to correctly guess the
upcoming required response with an accuracy of 70% before seeing predictable test
images. This might have reduced the task difficulty for predictable images and made
them less task relevant.

Our key hypothesis is that even perfectly orthogonalizing task relevance and
stimulus predictability could not preclude the proposed reduced attention effect for
predicable stimuli (as outline with the ABE). The reason for this is that the allocation of
attention towards a stimulus depends on both task relevance *and* stimulus saliency (as
visualized in Figure 1). We argue that it is impossible to experimentally orthogonalize
expectation and attention effects because unexpected stimuli are 'per definition' more
salient.

Why it is important to differentiate between a predictive coding and an
attention-based explanation of expectation suppression?

Over the last decades, theories of predictive coding became more and more popular
and expectation suppression is an observation that has frequently been interpreted as
supporting such theories. Although we agree that reduced prediction error coding
provides an appealing explanation for reduced responses to expected stimuli, we here
want to stress that predictive coding should not be accepted as an 'a-priori truth'.
Instead, scientist should remain open to considering alternative explanations in order
to prevent this theoretical framework from turning into a 'just-so story' that is able to
accommodate any empirical finding (Bowers & Davis, 2012; Heilbron & Chait, 2017).

In addition, differentiating between the two alternative explanations of
expectation suppression is important because they make different predictions about
what the increased neural activity for unexpected stimuli signifies: The PBE suggests
that this increased activity encodes the discrepancy between the actual and expected
input to a brain area, while the ABE suggests that this enhanced activity results from
increased attention allocation towards salient stimuli. As pointed out by Spratling
(2008), it is possible that stimulus saliency is computed in a similar manner as
prediction errors. However, even if stimulus saliency can be equated to a prediction
error coding process, we would like to point out that the ABE and the PBE differ with

regard to the order of attention effects and the encoding of prediction errors/stimulus saliency: The PBE proposes that attention facilitates prediction error coding (Feldman & Friston, 2010; Friston, 2009; Richter & de Lange, 2019; Spratling, 2008) while the ABE suggests that attention effects result from stimulus saliency (Desimone & Duncan, 1995, Kanan et al., 2009). Therefore, the critical difference appears to be that the PBE suggests that attention is *a cause* of the enhanced prediction-error coding while the ABE implies that attention is *an effect* of stimulus saliency.

Reduced prediction-error coding or a reduced attention effect: How to tell the difference and what is the more parsimonious explanation?

To differentiate between the predictive coding and an attention-based explanation of expectation suppression, we have to specify the different hypotheses that these explanations make and test for them empirically. As outlined above, the PBE and the ABE differ with regard to the relative timing of attention allocation and the timing of prediction error coding/stimulus saliency encoding. Therefore, future experiments could discriminate between the ABE and the PBE by using temporally resolved measurements (e.g., using EEG or MEG). For example, by testing the assumption of the PBE (Richter et al, 2019) that attention allocation precedes prediction error coding in the visual cortex (e.g., using an approach like Press et al., 2020).

How else could we determine whether expectation suppression results from predictive coding or from reduced attention to expected stimuli? As pointed out above (e.g., Figure 1), univariate responses cannot differentiate between these two possibilities, because PBE and ABE both predict reduced responses to expected stimuli (Alink et al., 2010; Blank & Davis, 2016). Also, the observation that stimulus predictability improves perceptual decisions (Kok et al., 2012; Schwiedrzik et al., 2007) does not enable us to rule out PBE or ABE, because they both are consistent with perceptual decisions being made based on sensory input *and* prior expectations - which in general leads to improved perceptual decisions for predictable stimuli. Since perceptual decision making is typically linked to brain activity in higher-level non-sensory brain areas (e.g., the frontoparietal network, Mulder et al., 2014), enhanced perceptual decision making for predictable stimuli does not necessarily require a facilitation of stimulus representations in sensory brain areas (Rungratsameetaweemana & Serences, 2019).

Could we disentangle the PBE from the ABE by investigating *how* features of expected and unexpected stimuli are encoded in the human brain – e.g., by analysing how expectation suppression affects fMRI stimulus decodability? Attention is known to enhance the responses of neurons tuned to the attended stimulus, leading to greater fMRI decodability (Kamitani & Tong, 2005). Therefore, according to the ABE, unexpected stimuli should be more decodable. In contrast, it is unclear whether according to the PBE unexpected or expected stimuli should be more decodable. If reduced responses to expected stimuli result from 'explaining away' of predicted sensory input (Clark, 2013) unexpected stimuli should be more decodable. However, stimulus predictability has been associated with enhanced fMRI pattern decodability which has led to the proposal that expectations give rise to sharpened neural representations (Kok et al., 2012). Such expectation-induced sharpening would contradict the ABE. At present, however, expectation-driven sharpening remains controversial because enhanced decodability has also been observed for unexpected stimuli (Jiang et al., 2013; Kumar et al., 2017; Richter et al., 2018; Tang et al., 2018) and given the observation that a conjunction of reduced fMRI responses and enhanced decodability can also be explained by a local-scaling effect (Alink et al., 2018). In conclusion, more evidence for sharpened responses to expected stimuli is needed to rule out the ABE.

Since the ABE and the PBE can both explain expectation suppression, one would ideally want to apply Ockham's razor to determine which provides the more parsimonious explanation (Heilbron & Chait, 2017). Such an approach, however, is challenging because attention and predictive coding models are typically not formulated at the same level of description (Marr, 1982). Therefore, resolving whether the ABE or the PBE is more parsimonious requires a reformulation of these explanations at a comparable level of description (e.g., see Spratling, 2008).

Can attention also explain other expectation effects?

Here, we discuss if predictive coding provides an alternative explanation of expectation suppression in comparison to traditional models of attention. Similar concerns may apply for other expectation-induced modulations of sensory responses, such as omission-evoked responses resembling responses to the expected stimulus (Bendixen et al., 2009) or feature-specific pre-stimulus activity induced by expectations (Fiser et

al., 2016; Kok et al., 2017). Although these observations are frequently proposed to result from predictive coding, increased neural activity in the absence of sensory stimulation has also been attributed to directed attention (Kastner et al., 1999).

Predictive coding, more than "New Labels for Old Ideas"?

A current debate around theories of predictive coding is whether they provide new constraints on brain function that have not already been provided by traditional models (Cao, 2020). In this perspective, we highlight that this applies to the expectation-suppression effect by pointing out that a traditional biased-competition model of attention can also explain reduced responses to expected stimuli. The current challenge for predictive coding is to conclusively show that it provides a more accurate description of how expectations suppress neural responses in sensory brain areas than traditional models.

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