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Dynamic contextual modulation in superior colliculus of awake mouse

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12		and agreed to be accountable for all aspects of the work in ensuring that questions related
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42 ABSTRACT

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The responses of neurons in the visual pathway depend on the context in which a stimulus is presented. Responses to predictable stimuli are usually suppressed, highlighting responses to unexpected stimuli that might be important for behaviour. Here we established how context modulates the response of neurons in the superior colliculus, a region important in orienting towards or away from visual stimuli. We made extracellular recordings from single-units in the superficial layers of superior colliculus in awake mice. We found strong suppression of visual response by spatial context (surround suppression) and temporal context (adaptation). Neurons showing stronger surround suppression also showed stronger adaptation effects. In neurons where it was present surround suppression was dynamic, and was reduced by adaptation. Adaptation's effects further revealed two components to surround suppression: one component that was weakly tuned for orientation and adaptable, and another component that was more strongly tuned but less adaptable. The selectivity of the tuned component was flexible, such that suppression was stronger when the stimulus over the surround matched that over the receptive field. Our results therefore reveal strong interactions between spatial and temporal context in regulating the flow of signals through mouse superior colliculus, and suggest the presence of a subpopulation of neurons that might signal novelty in either space or time.

1	SIGNIFICANCE STATEMENT
52	Our senses provide enormous amounts of information, and the central nervous system needs
53	to filter this information to focus on potentially important objects. Here we study two visual
54	mechanisms that might highlight unexpected or surprising objects for further analysis:
55	surround suppression and adaptation. We show that both mechanisms work to filter the neural
66	signals provided by the superior colliculus, a mid-brain area important for directing
57	behaviour. We also show that the two mechanisms are unexpectedly intertwined, endowing
8	rich dynamics on neural signals at the first central stage of sensory processing. Finally, our
59	results suggest a subpopulation of neurons that is specialised for signalling the presence of
0	potentially important objects.
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73 INTRODUCTION

74 Unexpected objects are likely to be important for behaviour, and predictable objects less 75 important. Many aspects of the functional organisation of the visual system can be explained 76 by supposing that neuronal activity is suppressed when the image falling on a receptive field 77 is predictable. For example, inhibitory inputs to retinal neurons can be thought of as providing 78 predictions about the intensity of the image over the receptive field, suppressing responses 79 unless the intensity deviates from those predictions (Srinivasan et al., 1982). The functional 80 consequences of this predictive inhibition are the classical centre-surround organisation and 81 transient responses of receptive fields in the retina and its targets (Hartline, 1940; Kuffler, 82 1953; Barlow, 2001). 83 In the classical model of an early receptive field, inhibition provides predictions about the 84 average intensity of the image over the receptive field, but not the variance, or pattern, of 85 intensity in that image. Two additional mechanisms are needed to explain how responses to 86 predictable patterns are suppressed. Spatial interactions (often called surround suppression) 87 can suppress responses when the pattern over the classical receptive field is similar to that in 88 the surrounding region. Temporal interactions (often called adaptation) can suppress 89 responses when the pattern is similar over time. The spatial and temporal suppression are 90 thought to reflect the action of 'gain controls', mechanisms that regulate the responses 91 generated by the classical receptive field (Solomon and Kohn, 2014; Webster, 2015; Shapley 92 and Victor, 1978; Bonds, 1989; Carandini and Heeger, 2011). 93 While most work on spatial and temporal gain controls has concentrated on visual cortex 94 (Allman et al., 1985; Solomon and Kohn, 2014; rodent: Adesnik et al., 2012; Vaiceliunaite et 95 al., 2013; Self et al., 2014; cat: Movshon and Lennie, 1979; Carandini and Ferster, 1997; 96 monkey: Mayo and Sommer, 2008; Patterson et al., 2013), gain controls are also known to be 97 important in the retina and early stages of central visual processing (rodent: Zhang et al.,

2012; Jacoby and Schwartz, 2017; lagomorphs: Oyster and Takahashi, 1975; Smirnakis et al.,
1997; cat: Sterling and Wickelgren, 1969; Jones et al., 2000; Bonin et al., 2005; Fisher et al.,
2017; monkey: Solomon et al., 2002; Solomon et al., 2004; Solomon et al., 2006; Boehnke et
al., 2011). In most animals the major target of the retina is the superficial layers of the mid-
brain superior colliculus (SC, homologous to the optic tectum) (May, 2006; Ellis et al., 2016).
The superficial layers of SC project to, among other areas, the deeper layers of SC, which are
important in organising movements towards or away from potentially important objects (Dean
et al., 1989; Basso and May, 2017; rodent: Comoli et al., 2012; Hoy et al., 2019). The
receptive fields of superficial SC neurons are often remarkably selective for image features:
for example, neurons in superficial SC of mouse can be tightly tuned for contour orientation,
even in the absence of visual cortex (Wang et al., 2010; Shi et al., 2017). The receptive fields
of neurons in superficial SC also show prominent surround suppression (rodent: Girman and
Lund, 2007; Wang et al., 2010; Ahmadlou et al., 2017; Barchini et al., 2018; monkey:
Davidson and Bender, 1991) and adaptation (Dutta and Gutfreund, 2014; monkey: Boehnke et
al., 2011;). How gain controls influence the response of SC neurons is less clear, particularly
in awake animals. For example, we do not know if surround suppression and adaptation's
effects are ubiquitous, whether they are independent, or how they interact.
Here we made extracellular recordings from the superficial layers of SC in awake mice. We
characterised surround suppression from the response to drifting gratings of varying size, and
characterised adaptation from the time-course of the response to drifting gratings of optimal
size. We find profound impact of surround suppression and adaptation in many but not all
neurons, and show that neurons with strong suppressive surrounds are also more susceptible
to adaptation. Further, the suppressive surrounds themselves are susceptible to adaptation, and
adaptation's effects reveal at least two components of suppression – an untuned component
that is adaptable, and a tuned component that is less adaptable. The selectivity of the tuned

component was not static, but flexible: that is, suppression depended on what was shown to
the receptive field, and was stronger when the stimulus over the surround matched that over
the receptive field. The spatial and temporal gain controls may therefore allow neurons in the
superficial SC the capacity to dynamically signal unexpected events in either space or time.

128	METHODS
129	Ethical approval.
130	All animal procedures were performed in accordance with the UK Animals Scientific
131	Procedures Act (1986). Experiments were performed at University College London in
132	accordance with its animal care committee's regulations, under personal and project licenses
133	released by the Home Office following appropriate ethics review, and in accordance with the
134	ethical policy under which eNeuro operates.
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136	General
137	Adult C57BL/6 male mice (8-12 weeks at the start of experiments, 20-35 g) were obtained
138	from Charles River Laboratories. Animals were housed with ad libitum food and water, on an
139	inverted 12-hour light/dark cycle. Measurements were obtained during the dark phase. To
140	prevent damage to implanted devices, animals were singly housed after the preparatory
141	surgeries described below.
142	Preparation for recordings: anaesthesia was induced with 3% isoflurane in O ₂ and the animal
143	transferred to a stereotaxic apparatus. Anaesthesia was subsequently maintained with 1-1.5%
144	isoflurane in O2, and adjusted as necessary by monitoring the breathing rate and absence of
145	reflex responses to paw pinch. The scalp was retracted and a craniotomy was made in one
146	hemisphere, centred 3.5-3.7 mm posterior to bregma, 0.7-1.1 mm lateral to the midline suture.
147	A metal head post fixed to the skull and a ground screw implanted over frontal cortex. In six
148	animals the brain was covered with a layer of Kwik-Cast Sealant (WPI), which was replaced
149	with artificial cerebrospinal fluid (Bio-techne Ltd, UK) during recording sessions; in these
150	cases, recordings were subsequently made using quartz/platinum-tungsten electrodes (Thomas
151	Recordings; impedance 4-5 M Ω) or tetrodes (impedance 0.5-0.8 M Ω). In two animals the
152	dura mater was instead removed and a 16-channel microdrive (arranged as 4 tetrodes, Axona

Ltd, UK) was implanted. Animals recovered from surgery for at least one week and were then
habituated to head-restraint before recordings started. Typical duration of a recording session
was 90-120 minutes. At the end of the experiments, animals were euthanized by overdose of
sodium pentobarbital I.P
Recordings and spike sorting: The analogue signal from each electrode was amplified and
filtered (0.3 kHz -7/10 kHz), then digitised and recorded at 48 or 44 kHz. All recordings
obtained at one site on one day were analysed together. Putative single-units were identified
off-line using Plexon Offline Sorter (Version 3.3.2, for single electrode recordings) or
KlustaSuite (Rossant et al., 2016). Single-units were identified by clustering in principal
component (PCA) space, followed by manual inspection of spike shape, auto- and cross-
correlograms. In no putative single unit did the fraction of ISIs under 0.5 ms exceed 2%.
Visual stimuli: Visual stimuli were generated using Expo (P. Lennie, Rochester, NY) on an
Apple Macintosh computer, and presented on a LCD monitor (Iiyama ProLite E1890SD,
mean luminance 35-45 candela/m ² ; 38 cm wide, 29 cm high) refreshed at 60 Hz and
displaying a grey screen of the mean luminance, positioned 20 cm from the animals' eye. The
monitor was gamma-corrected by measuring the luminance of the red, green and blue
elements with a photometer (Konica Minolta, Chroma meter CS-100A). Neural recordings
were aligned to the visual stimulus by the output of a photodiode scanning a small corner of
the stimulus monitor shielded from the animal. The coarse location of receptive fields was
manually identified and the monitor location adjusted to approximately centre them while
making the monitor normal to the animal. Receptive field position estimates were
subsequently refined by on-line analysis of responses to 'sparse-noise', where black or white
squares (size 15°; duration 0.2s) were presented pseudo-randomly at each location of a 9x9
grid centred in the monitor. Our recordings were made from a variety of elevations in the
nasal visual field, or from the lower temporal visual field, and were not distributed

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sufficiently for us to characterise the relationship between receptive field location and functional properties. We did not correct the display for the distortions in visual angle or changes in illumination that the short viewing distance produces at the edges of the monitor. Stimuli lasted for 2 s with an inter-stimulus interval of 0.5 s. Each set of stimuli included a blank condition (during which the screen was held at the mean luminance) from which 'spontaneous' or maintained firing rates were estimated. Each set of stimuli was presented in pseudo-randomised order for 3-15 repetitions. In some experiments we presented a drifting sinusoidal grating in a circular patch of varying diameter (2°-90°), outside of which the screen was held at the mean luminance. The spatial and temporal frequency of the gratings was determined by initial measurements at each site. We used a spatial frequency near the optimal for the neurons under consideration (usually 0.05 cycles/degree; μ 0.09, range 0.04-0.30); temporal frequency was usually 4 Hz (μ 3.5 Hz; 0.7 Hz, n = 2 units; 2 Hz, n = 33; 4 Hz, n = 56; 7.5 Hz, n = 6); Michelson contrast was 0.99(hereafter normalized to 1.0) unless varied. In additional experiments we presented a central patch of grating with a surrounding (abutting) annular grating. The central patch was of fixed size, and of the spatial and temporal frequency defined above; the annular grating was of the same spatial frequency, and a temporal frequency 0.5 Hz higher. In one experiment we varied the contrast of the annular grating, and in another experiment, we varied the orientation/direction of the annular grating. Each set of stimuli included trials in which the central patch or an annular grating was presented in isolation. Measurements were drawn from a large set of units, some of which have been reported previously (De Franceschi and Solomon, 2018). Data analysis Analysis. Offline analysis was performed in the Matlab environment (R2019a; Mathworks, Natick, MA). Peristimulus time histograms (PSTHs, bin width 0.016 s) were constructed for

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each trial, from which we extracted the mean firing rate. Unless stated, we define response as stimulus evoked activity, that is, the change in activity from that measured during presentation of a blank screen (the 'spontaneous' or maintained firing rate).

206 Inclusion criteria. We considered neurons visually responsive if their maximal response
207 exceeded the maintained rate by at least 1.5 S.D. of that rate, and further required that their
208 response exceed 2 impulses/s in the relevant analysis. We also required the centre of a units'
209 receptive field (estimated from responses to the sparse-noise stimulus) to be within 10° of the
210 stimulus centre.

Size tuning: To characterise the dependence of response on the size of a grating patch we assumed that both the classical receptive field and a suppressive surround could be described by concentric circular Gaussians (Cavanaugh et al., 2002a). The excitatory classical receptive field (*Le*) to a grating of diameter *d* is proportional to the integrated volume of a Gaussian:

$$L_e(d) = \frac{2}{\sqrt{\pi}} \int_0^d e^{-(x/r_e)^2} dx$$
 (1)

where r_e is the width of the Gaussian envelope. A similar expression can be derived for the larger surround Gaussian (Li). We assumed that the surround has divisive influence on the activity of the classical receptive field (Sceniak et al., 2001; Cavanaugh et al., 2002a), such that response is:

$$R(d) = \frac{K_e L_e(d)}{1 + K_i L_i(d)}$$
 (2)

Where K_e and K_i are respectively the excitatory and the suppressive gains. We found the set of parameters that maximised the log-likelihood (LL) of the model given the responses (El-Shamayleh & Movshon, 2011) using the Matlab function *fmincon*. We compared the model LL to an upper bound (LLu; obtained by fitting the responses to themselves) and a lower bound (LLl; obtained by fitting the responses to the average response across all stimuli). The

- normalized log-likelihood [*LLn* = (*LL-LLl*)/(*LLu-LLl*)] was used to decide whether to include
 the resulting model parameters in subsequent analyses (*LLn* >= 0.5). In addition to the
 parameters described above we included an additional parameter that allowed for a
 maintained discharge rate, and included in the set of responses to be modelled the activity
 during presentation of a blank grey screen. We estimated the preferred size from the model fit
 as the smallest size reaching 95% of the maximal response.
- 230 Suppression index: To quantify the suppression observed in size-tuning curves we calculated
- a suppression index (SI) as:

$$SI = 100 \times \frac{R_{opt} - R_{large}}{R_{opt}}$$
 (3)

- where R_{opt} is the response amplitude at the preferred size and R_{large} is the response amplitude
- at the largest tested size; both were extracted from the best predictions of the model above.
- We used the same expression to quantify suppression in centre-surround experiments,
- substituting R_{opt} with the response to a central patch alone, and R_{large} with the response to the
- relevant combination of central patch and annular grating.
- 237 Adaptation index. We calculated an adaptation index (AI) to characterise the change in
- response to a stimulus over time:

$$AI = 100 \times \frac{R_{early} - R_{late}}{R_{early}} \tag{4}$$

- where R_{early} and R_{late} are the average evoked activity during the first and last 0.5 s of stimulus
- presentation respectively (the stimuli lasted for 2 s).
- 241 Orientation/direction tuning: We calculated the direction tuning of stimulus-evoked responses
- or SI as the amplitude of the vector sum of responses or SI to different directions:

$$gDSI = \frac{\sum R_{\theta}e^{i\theta}}{\sum R_{\theta}}$$
 (5)

where R_{θ} is the response to a grating of direction θ . A global index of orientation selectivity is defined in the same way, but after doubling θ . The preferred direction or orientation is the angle of the relevant vector sum. We used the same expression to orientation/direction tuning of surround suppression by substituting R_{θ} with the suppression index (SI_{θ}) measured for an annular grating of direction θ .

248 Statistics

All statistical comparisons were performed in Matlab. Correlations are the Pearson's correlation coefficient, *r*. Statistical tests are Student's paired t-tests unless noted.

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252 **RESULTS** 253 Most models of receptive fields early in the visual pathway suppose that the signals of 254 different photoreceptors are given appropriate weight (which may be excitatory or inhibitory) 255 and then summed to provide a receptive field that drives spiking output. These models can be 256 used to characterise neurons with centre-surround receptive fields as well as those neurons 257 with more complex response properties, such as orientation tuning (for example: Cheong et 258 al., 2013). These models are, however, unable to explain why the response of neurons often 259 depends on the structure of the image beyond the receptive field, or the previous history of 260 stimulation. Explaining these dependencies requires supposing additional spatial and temporal 261 gain controls, which regulate the sensitivity of the receptive field. 262 The presence of spatial gain controls can be established by measuring the tuning of neurons to 263 the size of a pattern. We therefore varied the diameter of a patch of drifting grating that was 264 centred on the receptive field of the neuron under study (Fig 1A-D). The response of most 265 neurons was suppressed as the grating extended beyond the receptive field and into the 266 surrounding region, showing the presence of a spatial gain control, or suppressive surround. 267 The presence of temporal gain controls can be established from the time course of response to 268 a visual stimulus. All neurons responded robustly at the onset of a small stimulus. In some 269 neurons the response was sustained throughout the stimulus duration (Fig 1B,D), but in others 270 it was rapidly suppressed (Fig 1A,C). This adaptation effect shows the presence of a temporal 271 gain control. 272 Prevalence of spatial and temporal gain controls 273 We characterised the impact of the spatial gain control as the proportional reduction in 274 response to a large grating (a suppression index, or SI; Equation (3)). Here values of 0 275 indicate neurons in which there was no discernible suppression at large sizes, while values of 276 100 indicate neurons that only responded to small stimuli, and were completely suppressed by

larger ones. On average this suppression index was 76.9% (median 70.4, S.D. 147.9, n = 93;
Fig 2A-C) but there was substantial variability across the population of neurons. The absence
of suppression in some neurons might arise if their receptive fields are very large, and the gain
control was beyond the extent of the stimulus monitor. Our sample included neurons that
responded best to the largest grating we could produce and the filled bars in Fig 2C show that
neurons preferring large gratings (diameter greater than 30°) showed little suppression. Most
neurons, however, preferred gratings 10-30° in diameter (geometric mean 16.3°, median 15.6°,
n = 93) and in many of these neurons we saw little suppression even though the preferred size
was well within the monitor gamut. The measurements above were obtained for patterns of
high contrast. To establish the sensitivity of suppression in a sample of neurons we measured
the response to a patch of grating of optimal size, and varied the contrast of an annular grating
(not shown). In these neurons, suppression at 25% annulus contrast was on average 21.6%
(s.d. 15.7, $n = 12$), about half that at 100% annulus contrast (38.3, s.d. 22.0; $p = 0.0139$, paired
Students t-test). Spatial gain controls can therefore be engaged at low image contrast, and
their impact increases with contrast.
To characterise temporal gain control independently from spatial gain control we examined
responses to a patch of drifting grating of the preferred size for the neuron under study. We
measured the impact of temporal gain controls as the proportional reduction in response from
early (the first 0.5s) to late (the last 0.5s) time points, producing an adaptation index (AI;
Equation (4)) similar to the suppression index above. On average, later responses were
suppressed by 37.4% (median 36.0, S.D. 33.3, n = 98; Fig 2D-F) but as for surround
suppression we saw substantial variability across the population of neurons (Fig 2F). This
variability in AI was not explained by variation in temporal frequency of the grating (2 Hz: $\boldsymbol{\mu}$
33.1%, s.d. 23.8%, n = 33; 4 Hz: μ 38.4%, s.d. 36.1, n = 56). In a sample of neurons we
measured AI for a small patch of grating at low or high contrast (not shown): the adaptation

302	index at 25% contrast was on average 47.9% (s.d. 35.0, $n = 27$), if anything stronger than at
303	100% contrast (25.0%, s.d 72.1, n = 28; $p = 0.13$, paired Student's t-test). Thus temporal gain
304	controls are also sensitive to low image contrast.
305	The substantial variability in surround suppression and adaptation's effects raises the question
306	of whether the spatial and temporal gain controls are co-expressed in individual neurons. To
307	establish this, we compared the shape of the size-tuning curves for drifting gratings (provided
308	by the SI) and the time course of response for small patches of drifting grating (provided by
309	the AI). We found strong surround suppression in neurons that showed strong adaptation
310	effects (e.g. Fig 1A,C), and weak surround suppression in neurons that showed weak
311	adaptation effects (e.g. Fig 1B,D). Consequently, when we compared the index of surround
312	suppression (SI) and the index of adaptation (AI) we found a positive correlation (Fig 2G, $r =$
313	0.51, p < 0.00001, Pearson's correlation coefficient). Spatial and temporal gain controls
314	therefore appear to be co-expressed in individual neurons.
315	Many, but not all, neurons in SC are tuned for the orientation or motion direction of a grating.
316	We therefore asked if this tuning might predict the expression of surround suppression or
317	adaptation effects (not shown). We found little relationship between adaptation's effects (AI)
318	and global measures of orientation or direction tuning (respectively $r = 0.08, 0.16; p = 0.47,$
319	0.17; $n = 78$). We found more of a relationship for surround suppression (SI; respectively $r =$
320	0.33,0.36;p=0.0028,0.0014). Units with little surround suppression were usually weakly
321	tuned for orientation or direction, while units with strong surround suppression included units
322	with a range of tuning for orientation/direction.
323	Tuned and untuned contributions to spatial gain controls revealed by adaptation
324	Inspection of PSTHs for small and large stimuli showed that responses to small stimuli were
325	more transient – that is, adaptation's effects were stronger for small stimuli (Fig 3A). This

suggests that spatial and temporal gain controls interact in shaping neural response. We
characterised this interaction by generating size tuning curves for early and late responses. We
found less surround suppression at late time points (Fig 3B), and our index of suppression
consequently reduced over time (Fig 3C; on average from 69.0% to 51.2%, n = 73, p <
0.00001, paired Student's t-test). The reduced suppression at late time points suggests that
surround suppression is also adaptable.
If a landation about an about the constitution of assumption of assumption is made about a the tourists of
If adaptation changes the sensitivity of surround suppression it may also change the tuning of
surround suppression. Previous work shows that surround suppression in mouse SC can be
sensitive to the orientation and/or direction of a pattern (Ahmadlou et al., 2017; Barchini et
al., 2018). We confirmed that suppression in SC was usually strongest when the orientation
and direction of the annular grating matched that over the receptive field (Fig 4A,B). In many
neurons (for example, the unit in Fig 4B), and in the population average (Fig 4E), suppression
was similar for either direction of motion of a parallel annular grating. In other neurons
suppression was clearly stronger when the direction of the annular grating also matched that
in the central patch, and in others surround suppression was untuned. We therefore asked if
suppression was more tuned in neurons in which spiking response (Fig 4A,D) was also
strongly tuned. We used a global index of orientation or direction selectivity (see Methods) to
compare the tuning of neuronal responses to a single large grating, with the tuning of
suppression elicited by the annular gratings. In both cases, values of 0 indicate no tuning,
while values of 1 indicate spiking response or suppression for only one stimulus. There was
little correlation (Fig 4C, $r = -0.02/0.09$, $p = 0.90/0.53$, Pearson's correlation coefficient):
suppression was often tuned even when spiking response was untuned, and vice versa.
Nevertheless, in neurons where the tuning of suppression was strong enough to define a
preferred stimulus (tuning index greater than 0.1), the preferred orientation/direction of
suppression was generally aligned with the stimulus shown in the central patch (Fig 4F).

To establish if the tuning of suppression is changed by adaptation we measured the tuning of
suppression in the first 0.5s following the onset of the stimulus, and in the last 0.5s. The
population average showed strong suppression at early time points, but this suppression was
only weakly tuned for annulus orientation (Fig 5B). At later time points the overall strength of
suppression was reduced and was largely confined to gratings of the same orientation /
direction as the central patch, resulting in increased selectivity of suppression (Fig 5F). To
illustrate how tuning changed in individual neurons we compared suppression for annular
gratings of the same orientation and motion direction as the central patch, with that for
gratings tilted by 45° (average of +/- 45°; Fig 5C,G). Parallel gratings generated stronger
suppression at both time points, but their advantage was less pronounced at early (parallel
gratings generated μ 21.1% more suppression than tilted gratings, S.D. 22.3, n = 58; Fig 5C-
D) than late timepoints (μ 32.6%, S.D. 31.8, n = 36; Fig 5G-H) (p = 0.0059, paired Students
t-test, for 35 units that could be characterised at both time points). We conclude that the
overall strength of surround suppression reduces over time, and the selectivity of suppression
increases.
The pattern of results in Figs 5B&F might be explained if spatial gain controls constitute two
mechanisms – one that is narrowly tuned for orientation / direction and less susceptible to
adaptation, and one that is more broadly tuned and more susceptible to adaptation. One
potential source of suppression is the response of other neurons in SC and we therefore
conducted similar analyses of spiking response to large drifting gratings (Fig 5A,E). As for
suppression, the population spiking response has both tuned and untuned components, and as
for suppression the population spiking response reduced substantially at later time points,
showing the presence of adaptation effects. This reduction in response was similar for the
preferred grating and a grating tilted by 30° ($p = 0.0985$, $n = 34$). Similar results were
obtained if we compared responses to preferred and orthogonal gratings, examined direction

tuning curves in a larger dataset including additional units (n = 155; not shown), or compared
the global orientation and direction selectivity indices. We conclude that adaptation's effects
on spiking activity in SC may be sufficient to explain why the overall strength of suppression
is reduced at late time points, but additional mechanisms may be required to explain why the
tuning of suppression increases at late time points.
Flexible tuning selectivity of spatial gain controls
The tuning selectivity of spatial gain controls could either be static, or depend on the
parameters of the stimulus over the receptive field. In other words, the tuning selectivity may
be 'fixed' or 'flexible'. In a sample of neurons that were suppressed by annular gratings we
therefore repeated the measurements after rotating the orientation/direction of the central
patch by 45 degrees. Suppression is relatively broadly tuned, and we therefore expected to see
similar tuning curves for suppression across the two measurements. This was the case (Fig
6A,E). Nevertheless, the most suppressive surround did depend on the orientation/ direction
of the central patch, at both early (Fig 6A,B) and late (Fig 6E,F) time points.
To establish how surround suppression depended on what was shown over the receptive field
we focused our analyses on suppression evoked by the pair of annular gratings that matched
the orientations/directions of the pair of gratings shown to the receptive field. The orientations
of the central gratings over the receptive field are labelled C1 and C2 in Fig 6, and the annulus
orientations that matched them are respectively S1 and S2. The analyses in Fig 6C,G show
that annular gratings were relatively more effective when they matched the central patch. That
is, when C1 was the central grating, suppression at S1 was stronger than suppression at S2
(points lie below the diagonal in Fig 6C,G), and when C2 was the central grating the pattern
was reversed (points lie above the diagonal). To compare the suppression that was evoked by
S1 and S2 at each of the centre orientations we calculated the difference in suppression for the
two conditions - that is, for C1 we calculated SI _{C1} -SI _{C2} , and for C2 we also calculated SI _{C1} -

SI _{S2} . This subtraction collapsed the data along the diagonal in Fig 6C,G while preserving sign,
and produced the distributions in Fig 6D,H.
The advantage of matched annular gratings appeared to increase with time: suppression
indices lie further away from the diagonal at later time points (Fig 6G) than they do at early
time points (Fig 6C). Similarly, the distributions in Fig 6H lie further from 0 than do the
distributions in Fig 6D. To provide a statistical comparison across early and late time points,
we computed an additional index, [MI = $(SI_{C1,S1}-SI_{C1,S2}) - (SI_{C2,S1}-SI_{C2,S2})$] for each unit at
each time point. If suppression was fixed, and thus the same for any particular annulus
orientation, regardless of centre orientation, then this MI should be 0, but it was not (early: $\boldsymbol{\mu}$
40.4 , s.d. 33.1 , $n = 28$; late: μ 85.4 , s.d. 60.1 , $n = 22$). Comparison of the indices at early and
late time points, for units that responded in both, showed that the index increased at later time
points ($p = 0.000459$, n = 22; paired Students' t-test). We conclude that the tuning selectivity
of suppression in SC can be flexible, and that this flexibility is most apparent at later time
points, when the untuned gain control is less effective.

DISCUSSION

Functional impact of temporal gain controls
Adaptation's effects allow neurons to adjust their activity to the recent stimulus history. The
reduction in response to a stimulus that is unchanging might be used by neurons to better
signal changes over time. We have shown that many neurons in the mouse superior colliculus
show adaptation effects: responses are characterized by a large initial response that is quickly
suppressed. This suppression cannot be explained by retinal light adaptation, because we
presented drifting gratings, where the spatial pattern is constantly modulated.
The timescales of adaptation effects that we have characterised are on the order of 10s-100s of
milliseconds, shorter than most earlier characterisations of adaptation effects in SC, which
were primarily conducted in anaesthetised animals (e.g. rabbit: Horn and Hill, 1966; monkey:
Cynader and Berman, 1972; mouse: Drager and Hubel, 1975; pigeon: Woods and Frost,
1977). That work emphasised a large and long-lasting suppression of response (often called
habituation) by the appearance of a visual stimulus. The habituation was strongest in the
intermediate and deeper layers of the SC, but it has also been reported in the superficial layers
(rabbit: Oyster and Takahashi, 1975; cat: Binns and Salt, 1995). In anaesthetised rat, more
rapid adaptation effects were seen in superficial neurons for flashes of bright spots on a dark
background (Bytautiene and Baranauskas, 2018), but those effects may have included a
contribution of light adaptation. Our stimuli were interleaved, and were preceded by other sets
of stimuli, so it is difficult to establish the effect of long-term habituation from these
measurements. Nevertheless, we did not see a clear impact of position in the stimulus
sequence on response amplitude, even when we only considered units with high adaptation
indices (not shown). Recordings in superficial layers of awake monkey also show lack of
long-term adaptation effects (e.g. Goldberg and Wurtz, 1972), and more substantial short-term
adaptation effects (Mayo and Sommer 2008: Boehnke et al. 2011)

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Previous measurements of adaptation's effect in SC have often involved repeated presentation of a brief stimulus (e.g. Boehnke et al., 2011), whereas we measured response to a single, longer, continuous stimulus. While the two types of stimuli are likely to engage the same mechanisms, that does not mean they will have the same effect (Solomon and Kohn, 2014). The transients associated with repeated flashes may be more effective at driving the adaptive mechanism(s) and repeated presentations may therefore induce greater changes in activity. Alternatively, the periods of rest between the presentations may allow adaptive mechanisms to recover, and repeated presentations may therefore have less effect. Onset transients appear to be increasingly important for information processing as one ascends through the visual hierarchy (eg. Tovee et al., 1993; Muller et al., 2001), so differences in adaptive responses to repeated and continuous presentations may be more pronounced in later visual processing. SC integrates early and later visual inputs, so comparison of adaptive responses to flashed and continuous presentation may be of interest. Adaptation effects are prominent in retinal ganglion cell response, and likely first emerge in the bipolar cell input to ganglion cells (salamander: Chander and Chichilnisky, 2001; salamander/rabbit: Baccus and Meister, 2002; monkey: Solomon et al., 2004; guinea pig: Zaghloul et al., 2005; mouse: Marco et al., 2013). It is therefore probable that some of the adaptation effects that we see in SCs are inherited from the retinal input, but we are not aware of reports of retinal neurons that show the complete suppression of response that we often encountered in SCs. Additional mechanisms in SCs, potentially mediated by GABA_B receptors and metabotropic glutamate receptors, have been implicated in pre- and postsynaptic adaptation effects in SCs (cat: Binns and Salt, 1995; rat: Cirone and Salt, 2001), and these are likely to enhance or supersede adaptation effects inherited from retinal input. In addition, the sustained response was slightly reduced at large stimulus sizes (cf. Fig. 3A), while the initial transient was strongly reduced. Size-dependence of the sustained response

466	has also been observed in SC of monkey (Chen and Hafed, 2018), though direct comparison is
467	difficult because that work explored shorter time-windows and stimuli confined to the
468	receptive field.
469	Functional impact of spatial gain controls
470	Suppressive surrounds have been described in the superior colliculus of many species (cat:
471	Sterling and Wickelgren, 1969; monkey: Cynader and Berman, 1972; Wurtz et al., 1980; rat:
472	Girman and Lund, 2007; zebrafish: Del Bene et al., 2010; barn owl: Mysore et al., 2010;
473	Zahar et al., 2012; Zahar et al., 2018), including mouse (Wang et al., 2010; Gale and Murphy,
474	2014; Ahmadlou et al., 2017; Barchini et al., 2018). We show that in awake mouse that
475	surround suppression consists at least two components – one that is weakly tuned and
476	adaptable, and another that is more tuned and less susceptible to adaptation's effects. The
477	tuned gain controls appear to have flexible selectivity, such that the most suppressive
478	surrounding stimulus is that which matches the stimulus over the receptive field.
479	Most types of mouse retinal ganglion cell send axons to the SCs (Ellis et al., 2016) and
480	several of these are known to show surround suppression. One is the ON-OFF W3 cell (Zhang
481	et al., 2012), thought to be a homolog of the "net convexity detector" cells in the frog retina
482	(Lettvin et al., 1959) and the local edge detector (LED) cells first described in rabbit (Levick,
483	1967). But size sensitive responses are also found in 'high definition' (HD) (Jacoby and
484	Schwartz, 2017) as well as the direction-selective J and BD retinal ganglion cell classes,
485	which also project to SCs (Kim et al., 2010). Thus some of the surround suppression that we
486	observe in SCs may be inherited from the retinal input. Yet while adaptation effects can
487	reduce the amount of inhibition onto retinal ganglion cells (salamander/rabbit: Baccus and
488	Meister, 2002; mouse: Wark et al., 2009; Marco et al., 2013; salamander: Kastner et al.,
489	2019), we are not aware of reports of adaptation effects on suppression in retina at the time
490	scale of the rapid adaptation that we see in SCs. This suggests that lateral interactions within

SC	s are a strong contributor to the surround suppression that we see, and the simplest
con	aclusion is that adaptation reduces surround suppression in SCs because adaptation reduces
spil	king activity in SCs.
Net	tworks in the superficial layers of mouse SC include inhibitory lateral interactions that
sup	press the activity of simultaneously activated neurons (Phongphanphanee et al., 2014).
Loc	cal inhibition from 'horizontal cells', which respond to large stimuli (Gale and Murphy,
201	(4), may be particularly important in providing surround suppression (Gale and Murphy,
201	6), while 'narrow field' and 'wide field' cells appear particularly susceptible to
sup	pression (Gale and Murphy, 2014). Similar mechanisms for constructing size tuning have
bee	en described in the zebrafish optic tectum (Del Bene et al., 2010). In addition, the superior
col	liculus receives substantial input from visual cortex (May, 2006), though the role(s) of
cor	tico-collicular input remain unclear – these projections modulate gain of SC neurons but
the	ir absence seems to have little effect on tuning properties (Wang et al., 2010; Zhao et al.,
201	4) or surround suppression (Ahmadlou et al., 2017), at least in mouse. Indeed, surround
sup	pression in the SC may precede that in V1 (monkey: White et al., 2017) and inactivation
of S	SC can interfere with surround suppression in V1 of mouse (Ahmadlou et al., 2018).
Un	its that were not selective for pattern orientation/direction were also less likely to show
stro	ong surround suppression. This result may reflect a straightforward correlation in the two
fun	ctional properties, or surround suppression may be important for constructing selectivity
for	orientation or direction. Regardless, controlling for the size of stimuli is likely to be
imp	portant in characterising, and therefore understanding, the mechanisms of orientation and
dire	ection tuning in SC.
We	e found that tuned surround suppression was less susceptible to adaptation than untuned
sup	pression, with the consequence that suppression was more sharply tuned and more flexible
in 1	ater activity. Our finding that at least some of the suppression in SCs is flexible is in

accord with recent calcium imaging from SCs of anaesthetised mouse (Barchini et al., 2018).
That work showed suppression by surrounding gratings of the same motion direction as a
central patch, and facilitation by surrounds of the opposite direction, particularly in excitatory
cells. The dynamics of calcium signalling make comparison of response time course difficult,
but the initial spiking response, where we find weakly tuned suppression, may have
contributed less to the calcium signal than the subsequent response, where we find more
tuning of suppression and some facilitation. Our finding that tilted surrounds could even
become facilitatory in the late phase of responses raises the possibility that the tuning of late
suppression may in fact reflect tuned facilitation. Flexible suppression selectivity may
therefore reflect input from neurons with large receptive fields that are sensitive to image
continuity (if they provide tuned suppression) or sensitive to image discontinuity (if they
provide tuned facilitation). These flexible mechanisms may arise in SCs or in its inputs. If
they arise in SCs, then one candidate may be the horizontal cells. Regardless, mouse SCs is
likely to be a useful model for understanding the mechanisms that enable flexible suppression
of neural responses by spatial context (Coen-Cagli et al., 2015).
The functional properties of surround suppression in SCs are remarkably similar to that
described for primary visual cortex (V1) in many mammals. First, surround suppression in V1
is often orientation- and direction selective (mouse: Self et al., 2014; cat: Nelson and Frost,
1978; DeAngelis et al., 1994; Ozeki et al., 2004; monkey: Sillito et al., 1995; Levitt and Lund,
1997; Cavanaugh et al., 2002b; Webb et al., 2005; Henry et al., 2013), and that tuning
selectivity can be flexible (Sillito et al., 1995; Cavanaugh et al., 2002b). Second, in V1 of
mouse, monkey and human, this tuned suppression is complemented by an untuned
suppression (mouse: Self et al., 2014; monkey: Webb et al., 2005; Henry et al., 2013; human:
Schallmo et al., 2019), some of which may be inherited from earlier processing (cat/monkey:
Sillito et al., 1993; cat: Ozeki et al., 2004; Bonin et al., 2005; Naito et al., 2007; monkey:

541 Solomon et al., 2002; Camp et al., 2009). Third, some components of surround suppression in 542 V1 of monkey and human are susceptible to adaptation (Cavanaugh et al., 2002a; Wissig and 543 Kohn, 2012; Patterson et al., 2013; Schallmo and Murray, 2016), though in monkey V1 the 544 tuned components of suppression may be more sensitive to adaptation than the untuned 545 components (Webb et al., 2005). 546 Summary 547 We have shown the presence of spatial and temporal gain controls in SCs of awake mouse 548 and how they are distributed across neurons. Our results are consistent with the idea that these 549 gain controls provide a predictive signal against which activation of the classical receptive 550 field is compared, thereby suppressing the response to predictable stimuli and highlighting 551 unexpected ones. Our results can be accommodated by a layering of gain controls as 552 illustrated in Fig 7. Fig 7A shows the standard model of early visual processing (e.g. 553 Carandini and Heeger, 2011; Solomon and Kohn, 2014). The output of the classical receptive 554 field (CRF), which filters the visual image, is subject to a spatial gain control, or suppressive 555 surround, before driving spiking activity. The suppressive surround is constructed from 556 nearby neurons with similar characteristics. Adaptation's effects can be thought of as 557 changing the output function of the neuron, as shown by the red-line in Fig 7B. This accounts 558 for the results in Fig 5, because similar neurons contribute to the surround, and the surround is 559 therefore relatively broadly tuned for orientation/direction and susceptible to adaptation's 560 effects. To account for the advantage of surrounds that match the centre stimulus (Fig 6), and 561 the apparent resilience of this suppression to adaptation's effects, a second mechanism seems 562 to be required (Fig 7C). This is sensitive to the relationship between features over the CRF 563 and surround, and is less adaptable. 564 We also found that the strength of adaptation's effects and strength of surround suppression 565 were correlated among neurons. Our results therefore show that neurons characterized by a

transient, adapting response are more likely to also be affected by spatial context, and may
therefore signal the presence of unexpected objects in either the spatial and the temporal
domain. This suggests the presence of a subpopulation of 'novelty' or 'saliency' neurons
within the SC that are sensitive to unexpected events in the visual diet. Whether this
functional subgroup has an anatomical correlate would be of interest. In monkey, the
amplitude of the initial transient response in anatomically deeper visual-motor SC neurons,
which receive direct input from the superficial purely visual neurons studied here, is known to
be particularly important in the nature and latency of orienting behaviours such as saccades
(Boehnke and Munoz, 2008; Chen and Hafed, 2017).

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Figure 1. Expression of spatial and temporal gain controls in neurons in superior colliculus of awake mouse. A-D. Responses of four representative neurons. The left panel in each case shows the average firing rate of the neuron during a 2 s presentation of a circular patch of drifting grating at a spatial frequency near the preferred for the neuron $(0.05\text{-}0.07\text{ cycles/}^\circ)$, and centred on the receptive field. The right panel in each case shows the peristimulus time histogram (PSTH, bin width 0.016 s) during presentation of a patch of grating near the preferred size for that neuron, which is indicated by the arrow in the left panel. Dashed horizontal lines show the maintained rate in absence of patterned visual stimulus. Solid line shows the predictions of the size-tuning model described in Methods (Equation (2)). Error bars show ± 1 SEM across trials. The insets show a spatial map of responses (white indicates no activity and darker colours indicate stronger responses) to a black square, 15° wide, flashed at each of 81 positions on the monitor. Calibration bars are 30° . The schematic below panel (A) shows the relative size of two patches of grating, and how a grating of 0.05 cycles/ $^\circ$ would appear in each of them.

Figure 2. Correlated variability in surround suppression and adaptation effects. A. Population size-tuning for patches of drifting gratings. Each row of the image shows the predictions of the size-tuning model for a single neuron (as in Fig 1), normalized to its maximum response. Only units in which the normalized log-likelihood of the model was at least 0.5 are shown. The units are ordered, from bottom-to-top, by the preferred size. B. Mean size-tuning for a drifting grating, obtained by averaging across the rows in (A). Dashed lines show ±1 SEM across neurons. Dashed horizontal line shows the maintained rate, normalized to the unit's maximum visual response before averaging. Arrow indicates the definition of the suppression index, SI, which is the proportional reduction in response from a

grating of preferred size to a large grating. **C**. Distribution of the suppression index across the population of units in (A). The filled bars show SI for neurons with preferred diameter greater than 30°. **D**. Population time-course for drifting gratings of preferred size. The units are ordered, from bottom-to-top, by the adaptation index, AI. Colour bar as in (A). **E**. Mean time-course for a drifting grating, obtained by averaging across the rows in (D). Conventions as in (B). Arrow indicates the definition of the adaptation index, AI, which is the proportional reduction in response from the first 0.5s to the last 0.5s. **F**. Distribution of the adaptation index across the population of units in (D). **G**. Comparison of the suppression index and adaptation index in individual neurons (n = 93). Dashed line is the best linear fit to the data.

Figure 3. Surround suppression is susceptible to adaptation. A. Time-course of population response for gratings for small (20° diameter) and large (90°) patches of drifting grating. Responses were normalized to the mean response across all patch sizes (most of which are not shown) before averaging (n = 98). Dashed horizontal line shows the maintained rate in absence of patterned visual stimulus, normalized in the same way. Error bars are omitted for clarity. B. Size-tuning for early (0-0.5 s) and late (1.5-2 s) response, normalized in the same way as (A).. Error bars are ± 1 SEM across neurons that passed criteria for inclusion (early: n = 92; late: n = 80). C. Comparison of suppression index for early and late response (n = 73). Dashed line shows the unity line. Points falling below the line indicate neurons in which suppression was stronger in the early response than in the late response.

Figure 4. Tuning of surround suppression in SC of awake mouse. A,B. Response of an example neuron. **A.** Tuning of spiking activity evoked by a large patch of drifting grating (45° diameter) of varying orientation/direction. Dashed horizontal line shows the maintained rate in absence of patterned visual stimulus. **B.** Tuning of suppression induced by an annular grating of varying orientation/direction. Responses are shown for presentation of a 15° patch

of drifting grating (upper dashed line, 'Centre alone') of direction 180°, the same stimulus when abutted by an annular grating of outer diameter 80° ('Centre+surround'), and two of the annular gratings presented in absence of the centre grating ('Surround alone'). The lower dashed horizontal line shows the maintained rate in absence of patterned visual stimulus. A suppression index, SI, can be calculated for each annulus direction as the proportional reduction in response from the 'Centre alone' stimulus to the relevant 'Centre+surround' stimulus. Error bars in (A,B) are ± 1 SEM over trials. C. Comparison of tuning for spiking activity (abscissa) and suppression (ordinate). Each unit contributes two points: the open symbols indicate a global measure of direction tuning (Equation (5)) and the filled symbols indicate a similar measure of orientation tuning. D, E. Population averages. D. Population average spiking activity evoked by a large grating, after aligning each neuron to its preferred direction, subtracting the maintained rate, and then normalizing by the mean response across all stimuli. E. Average suppression index, obtained as in (B), after aligning each neuron to the direction of the central grating patch. Dashed horizontal line shows an SI of zero. Error bars in (D,E) are ±1 SEM over neurons. F. Distribution of preferred orientation of suppression, relative to the orientation of the centre grating, in units in which the preferred orientation could be defined (orientation tuning index greater than 0.1). A relative orientation of zero indicates neurons in which the most suppressive stimulus was the same orientation as the centre; a relative orientation of 90 indicates neurons in which the most suppressive stimulus was orthogonal to the central stimulus. Schematics in panels A,D,E are not to scale.

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Figure 5. Impact of adaptation on tuning of spiking activity and surround suppression.

A-D. Responses in early (first 0.5s) time points. **A.** Population average tuning of spiking activity evoked by a large grating. Conventions as in Fig 4D. **B.** Population average tuning of suppression induced by annular grating. Conventions as in Fig 4E. **C.** Comparison of suppression index for annular gratings that match the direction of the central grating

('parallel', abscissa), or are tilted by 45° (ordinate). Suppression index for the latter was averaged across both possible directions of tilt. Dashed lines are the unity line. **D**. Distribution of the difference in suppression for the two annular gratings. Positive indices indicate more suppression by a parallel annulus than a tilted annulus. **E-H**. Same A-D, but for late (last 0.5s) time points. Schematics in panels A–C are not to scale.

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Figure 6. Adaptation's effects magnify flexible surround suppression. A-D. Responses in early (first 0.5s) time points. A. Spiking activity of an example neuron during presentation of a central stimulus either presented alone (C1,C2), or abutting an annulus of varying direction. Two measurements were made, first with a central grating near the preferred direction of spiking activity (C1) and then with a central grating tilted by 45° (C2). Error bars are ± 1 SEM over trials. B. Population average tuning of suppression for each of two central gratings (C1,C2). Conventions as in Fig 4E. C. Comparison of suppression induced by pairs of stimuli. The abscissa shows the suppression induced by an annular grating (S1) that matched the direction of the central grating C1. The ordinate shows the suppression induced by an annular grating (S2) that matched the central grating C2. Points below the unity line (dashed line) indicate stronger suppression for S1 than S2. Filled circles show measurements obtained with C1: these generally lie below the unity line, indicating that when C1 is used, S1 is more effective than S2. Open circles show measurements obtained for C2: measurements are generally above the unity line, indicating that S2 is more effective than S1. This is the pattern of results expected if suppression is stronger when the stimuli over the centre and surround are matched. **D.** Distribution of the difference in suppression for each of the centre gratings, C1 (lower) and C2 (upper). Positive indices indicate more suppression by S1 than S2. E-H. Same as A-D but for late (last 0.5s) time points. Example neuron in E is the same as that in A. Schematics above A,E are not to scale.

Figure 7. Descriptive model for interaction of spatial and temporal gain controls in SCs.
A. Standard model of receptive fields early in the visual pathway. The classical receptive field
('CRF') filters the visual image, and its output is subject to a spatial gain control ('Surround')
before driving spiking output. The surround is comprised of units with receptive fields similar
to that of the CRF. B. Adaptation's effects reduce the response of the neuron under study, and
the responses of neurons that contribute to the spatial gain control. C. Addition of a second,
less adaptable, component to the spatial gain control allows for preservation of suppression
when the features of the image over the CRF matches that over the surround.

















