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1 A century of vision research has demonstrated that symmetry contributes to numerous domains of visual perception (? ? ? ?). In
2 a 2-D image, the four fundamental symmetries, reflection, rotation,
3 translation and glide reflection, can be combined in 17 distinct ways.
4 These 17 “wallpaper” groups (? ? ?) obey a hierarchy, determined
5 by mathematical group theory, where simpler groups are subgroups
6 of more complex ones(?). Here we probe representations of sym-
7 metries in wallpaper groups using two methods: Steady-State Visual
8 Evoked Potentials (SSVEPs) recorded using EEG and psychophysical
9 detection thresholds. We find that the hierarchical relationships
10 between the wallpaper groups are almost perfectly preserved in both
11 behavior and response amplitudes in visual cortex. This remarkable
12 consistency between the structure of symmetry representations and
13 mathematical group theory, is likely generated over visual develop-
14 ment, through implicit learning of regularities in the environment.

Keyword 1 | Keyword 2 | Keyword 3 | ...

1 Symmetries are present at many scales in images of natu-
2 ral scenes, due to a complex interplay of physical forces
3 that govern pattern formation in nature. The importance of
4 symmetry for visual perception has been known at least since
5 the gestalt movement of the early 20th century. Since then,
6 symmetry has been shown to contribute to the perception of
7 shapes (? ?), scenes (?) and surface properties (?), as well
8 as the social process of mate selection (?). Most of this work
9 has focused on mirror symmetry,

10 In the two spatial dimensions relevant for images, the four
11 fundamental symmetries, reflection, rotation, translation and
12 glide reflection, can be combined in 17 distinct ways, the
13 “wallpaper” groups (? ? ?). The 17 wallpaper groups
14 obey a hierarchy of complexity, determined by mathematical
15 group theory, where simpler groups are sub-groups of more
16 complex ones (?). Here we use Steady-State Visual Evoked
17 Potentials (SSVEPs) to measure responses to 16 out of 17
18 wallpaper groups and show that activity in human visual cortex
19 is remarkably consistent with the hierarchical relationships
20 between the wallpaper groups. Specifically, the amplitudes of
21 symmetry-specific responses in individual participants ($n = 25$)
22 preserve these relationships at an above-chance level in 88.3%
23 (53 out of 60) of cases. Visual cortex thus encodes all the
24 fundamental symmetries using a representational structure
25 that closely approximates the subgroup relationships from
26 group theory. Given that most participants had no knowledge
27 of group theory, the ordered structure of visual responses to
28 wallpaper groups is likely learned implicitly from regularities
29 in the visual environment.

Results

The visual stimuli for our experiment were multiple exemplar images belonging to each of the 17 wallpaper groups, generated from random-noise textures, as described in detail elsewhere (?). To isolate brain activity specific to the symmetry structure in the images from activity associated with modulation of low-level features, we used a steady-state design, in which exemplar images belonging to 16 of the 17 wallpaper groups alternated with phase-scrambled images of the same group. Because all wallpapers are periodic images due to their lattice tiling structure, the phase-scrambled images are also a wallpaper group (P1). P1 contains no symmetries other than translation, while all other groups contain translation in combination with one or more of the other three fundamental symmetries (reflection, rotation, glide reflection) (?). Exemplars from each of the 16 groups alternated at 0.83 Hz with their corresponding set of P1 exemplars, that were matched in terms of their Fourier power spectrum. Because the P1 group serves a control stimulus in this approach, the experiment was restricted to the remaining 16 groups. This design allows us to isolate responses to structural features beyond the shared power spectrum, including any symmetries other than translation, in the odd harmonics of the image update frequency (? ?) [NORCIA 2002 MISSING]. Thus, the magnitude of the odd harmonic response components can be used as a distance metric for each group, with distance being measured relative to the simplest group, P1. A wallpaper group is a topologically discrete group of isometries of the Euclidean plane, i.e. transformations that preserve distance (?). Wallpaper groups differ in the number and kind of these transformations. In mathematical group theory, when the elements of one group is completely contained in another, the inner group is called a subgroup of the outer group (?). Subgroup relationships between wallpaper groups can be distinguished by their indices. The index of a subgroup relationship is the number of cosets, i.e. the number of times

Significance Statement

Authors must submit a 120-word maximum statement about the significance of their research paper written at a level understandable to an undergraduate educated scientist outside their field of speciality. The primary goal of the Significance Statement is to explain the relevance of the work in broad context to a broad readership. The Significance Statement appears in the paper itself and is required for all research papers.

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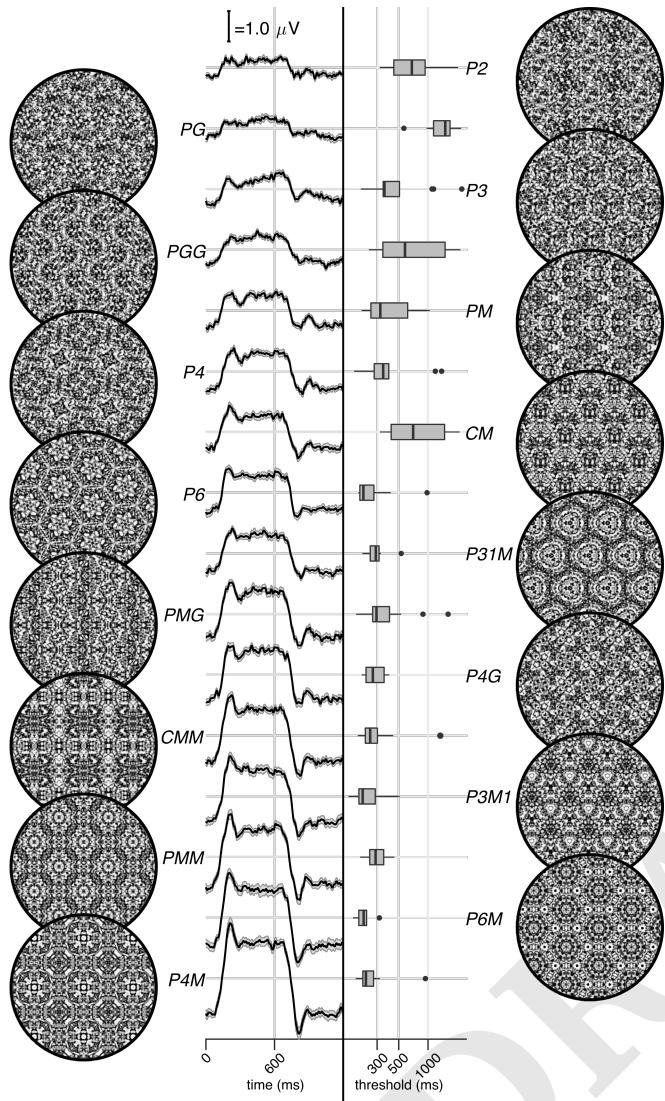


Fig. 1. Posterior distributions for the difference in mean RMS EEG response. Colour coding of the text indicates the index of the subgroup, while the colour of the filled distribution relates to the conditional probability that the difference in means is greater than zero. We can see that 55/64 subgroup relationships have $p(\Delta|data) > 0.9$.

the subgroup is found in the outer group (?). As an example, let us consider groups P6 and P2. If we ignore the translations in two directions that both groups share, group P6 consists of the set of rotations $0^\circ, 60^\circ, 120^\circ, 180^\circ, 240^\circ, 300^\circ$, in which P2 $0^\circ, 180^\circ$ is contained. P2 is thus a subgroup of P6, and the full P6 set can be generated by every combination of P2 and rotations $0^\circ, 120^\circ, 240^\circ$. Because P2 is repeated three times in P6, P2 is a subgroup of P6 with index 3 (?).

Discussion

Here we show that beyond merely responding to the elementary symmetry operations of reflection (?) and rotation (?), the visual system explicitly represents hierarchical structure of the 17 wallpaper groups, and thus the compositions of all four of the fundamental symmetry transformations (rotation, reflection, translation, glide reflection) which comprise regular textures. The RMS measure of SSVEP amplitude, preserves the complex hierarchy of subgroup relationships among the

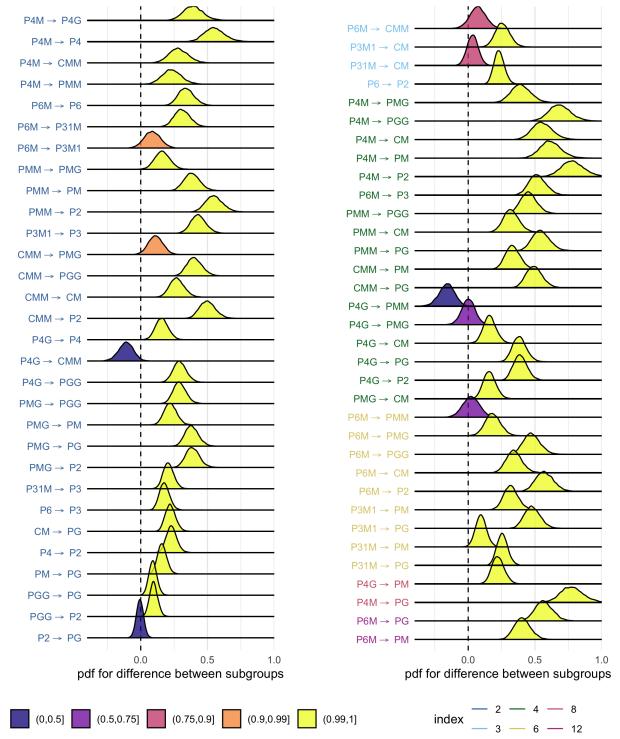


Fig. 2. Posterior distributions for the difference in mean RMS EEG response. Colour coding of the text indicates the index of the subgroup, while the colour of the filled distribution relates to the conditional probability that the difference in means is greater than zero. We can see that 55/64 subgroup relationships have $p(\Delta|data) > 0.9$.

82 wallpaper groups (?). Out of a total of 60 relationships, 53₃
 83 were preserved in a significant number of participants, and 49₄
 84 were significant even at a stricter threshold ($p < 0.002$). The₅
 85 ordering was highly stable in individual participants, with an₆
 86 average preservation rate of 21 of 25 participants across all 60₇
 87 relationships (see Figure 3). This remarkable consistency was₈
 88 specific to the odd harmonics of the stimulus frequency, that₉
 89 capture the symmetry-specific response (?) and to electrodes₁₀
 90 in an ROI over occipital cortex. When the same analysis was₁₁
 91 done on the even harmonics of the occipital cortex ROI, the
 92 ordering of responses was much less apparent (see Figure S2) and
 93 preservation rates much lower (see Figure S4). The odd har-
 94 monics from electrodes in an ROI over parietal cortex, showed
 95 even weaker evidence of preserving the hierarchy among sub-
 96 groups (see Figure S5). Importantly, no relationships were
 97 preserved in either of these control analyses that were not also
 98 preserved in the main analysis of the odd harmonics in the
 99 occipital cortex ROI. The current data provide a complete
 100 description of the visual system's response to symmetries in
 101 the 2-D plane. Our design does not allow us to independently
 102 measure the response to P1, but because each of the 16 other
 103 groups produce non-zero odd harmonic amplitudes (see Figure
 104 2), we can conclude that the relationships between P1 and all
 105 other groups, where P1 is the subgroup, are also preserved by
 106 the visual system. The subgroup relationships are not ob-
 107 vious perceptually, and most participants had no knowledge of
 108 group theory. Thus, the visual system's ability to preserve the
 109 subgroup hierarchy does not depend on explicit knowledge of
 110 the relationships. Furthermore, behavioral experiments have
 111 shown that although naïve observers can distinguish many of
 112 the wallpaper groups (?), they are generally error-prone when
 113 asked to assign exemplar images to the appropriate group (?).
 114 The correspondence between responses in the visual system
 115 and group theory that we demonstrate here, may reflect a
 116 form of implicit learning that depends on the structure of the
 117 natural world. The environment is itself constrained by phys-
 118 ical forces underlying pattern formation and these forces are
 119 subject to multiple symmetry constraints (?). The ordered
 120 structure of responses to wallpaper groups could be driven by
 121 a central tenet of neural coding, that of efficiency. If coding is
 122 to be efficient, neural resources should be distributed in such
 123 a way that the structure of the environment is captured with
 124 minimum redundancy considering the visual geometric optics,
 125 the capabilities of the subsequent neural coding stages and
 126 the behavioral goals of the organism (? ? ? ?). Early work
 127 within the efficient coding framework suggested that natural
 128 images had a $1/f$ spectrum and that the corresponding re-
 129 dundancy between pixels in natural images could be coded
 130 efficiently with a sparse set of oriented filter responses, such as
 131 those present in the early visual pathway (? ?). Our results
 132 suggest that the principle of efficient coding extends to a much
 133 higher level of structural redundancy – that of symmetries in
 134 visual images. The 17 wallpaper groups are completely regular,
 135 and relatively rare in the visual environment, especially when
 136 considering distortions due to perspective and occlusion. Near-
 137 regular textures, however abound in the visual world, and
 138 can be approximated as deformed versions of the wallpaper
 139 groups (?). The correspondence between brain data and
 140 group theory demonstrated here may indicate that the visual
 141 system represents visual textures using a similar scheme, with
 142 the wallpaper groups serving as anchor points in representa-

tional space. This framework resembles norm-based encoding
 143 strategies that have been proposed for other stimulus classes,
 144 most notably faces (?), and leads to the prediction that
 145 adaptation to wallpaper patterns should distort perception
 146 of near-regular textures, similar to the aftereffects found for
 147 faces (?). Field biologist have demonstrated that animals
 148 respond more strongly to exaggerated versions of a learned
 149 stimulus, referred to as “supernormal” stimuli (?). In the
 150 norm-based encoding framework, wallpaper groups can be
 151 considered super-textures, exaggerated examples of the near-
 152 regular textures that surround us. Artists may consciously
 153 or unconsciously create supernormal stimuli, to capture the
 154 essence of the subject and evoke strong responses in the audi-
 155 ence (?). Wallpaper groups are visually compelling and have
 156 been widely used in human artistic expression going back to
 157 the Neolithic age (?). If wallpapers are super-textures, their
 158 prevalence may be a direct consequence of the strategy the
 159 human visual system uses for encoding visual textures.
 160

Participants. Twenty-five participants (11 females, mean age
 161 28.7±13.3) took part in the EEG experiment. Their informed
 162 consent was obtained before the experiment under a proto-
 163 col that was approved by the Institutional Review Board of
 164 Stanford University. 11 participants (8 females mean age
 165 20.73±1.21) took part in the psychophysics experiment. All
 166 participants had normal or corrected-to-normal vision. Their
 167 informed consent was obtained before the experiment under a
 168 protocol that was approved by the University of Essex's Ethics
 169 Committee.
 170

Stimulus Generation. Exemplars from the different wallpaper
 171 groups were generated using a modified version of the method-
 172 ology developed by Clarke and colleagues(?) that we have
 173 described in detail elsewhere(?). Briefly, exemplar patterns
 174 for each group were generated from random-noise textures,
 175 which were then repeated and transformed to cover the plane,
 176 according to the symmetry axes and geometric lattice spe-
 177 cific to each group. The use of noise textures as the starting
 178 point for stimulus generation allowed the creation of an al-
 179 most infinite number of distinct exemplars of each wallpaper
 180 group. For each exemplar image, phase-randomized control
 181 exemplars were generated that had the same power spectrum
 182 as the exemplar images for each group. The phase scrambling
 183 eliminates rotation, reflection and glide-reflection symmetries
 184 within each exemplar, but the phase-scrambled images inher-
 185 ent the spectral periodicity arising from the periodic tiling.
 186 This means that all control exemplars, regardless of which
 187 wallpaper group they are derived from, degenerate to another
 188 symmetry group, namely P1. P1 is the simplest of the wallpa-
 189 per groups, and contains only translations of a region whose
 190 shape derives from the lattice. Because the different wallpaper
 191 groups have different lattices, P1 controls matched to different
 192 groups have different power spectra. Our experimental design
 193 takes these differences into account by comparing the neural
 194 responses evoked by each wallpaper group to responses evoked
 195 by the matched control exemplars.
 196

Stimulus Presentation. Stimulus Presentation. For the EEG
 197 experiment, the stimuli were shown on a 24.5" Sony Trimmerster
 198 EL PVM-2541 organic light emitting diode (OLED) display
 199 at a screen resolution of 1920×1080 pixels, 8-bit color depth
 200 and a refresh rate of 60 Hz, viewed at a distance of 70 cm.
 201

The mean luminance was 69.93 cd/m² and contrast was 95%. The diameter of the circular aperture in which the wallpaper pattern appeared was 13.8° of visual angle presented against a mean luminance gray background. Stimulus presentation was controlled using in-house software.

were band-pass filtered from 0.3 to 50 Hz. Raw data were evaluated off line according to a sample-by-sample thresholding procedure to remove noisy sensors that were replaced by the average of the six nearest spatial neighbors. On average, less than 5% of the electrodes were substituted; these electrodes were mainly located near the forehead or the ears. The substitutions can be expected to have a negligible impact on our results, as the majority of our signal can be expected to come from electrodes over occipital, temporal and parietal cortices. After this operation, the waveforms were re-referenced to the common average of all the sensors. The data from each 12s trial were segmented into five 2.4 s long epochs (i.e., each of these epochs was exactly 2 cycles of image modulation). Epochs for which a large percentage of data samples exceeding a noise threshold (depending on the participant and ranging between 25 and 50 μ V) were excluded from the analysis on a sensor-by-sensor basis. This was typically the case for epochs containing artifacts, such as blinks or eye movements. The use of steady-state stimulation drives cortical responses at specific frequencies directly tied to the stimulus frequency. It is thus appropriate to quantify these responses in terms of both phase and amplitude. Therefore, a Fourier analysis was applied on every remaining epoch using a discrete Fourier transform with a rectangular window. The use of epochs two-cycles (i.e., 2.4 s) long, was motivated by the need to have a relatively high resolution in the frequency domain, $\delta f = 0.42$ Hz. For each frequency bin, the complex-valued Fourier coefficients were then averaged across all epochs within each trial. Each participant did two sessions of 8 trials per condition, which resulted in a total of 16 trials per condition.

EEG Analysis. Response waveforms were generated for each group by selective filtering in the frequency domain. For each participant, the average Fourier coefficients from the two sessions were averaged over trials and sessions. The Steady-State Visual Evoked Potentials paradigm we used allowed us to separate symmetry-related responses from non-specific contrast transient responses. Previous work has demonstrated that symmetry-related responses are predominantly found in the odd harmonics of the stimulus frequency, whereas the even harmonics consist mainly of responses unrelated to symmetry, that arise from the contrast change associated with the appearance of the second image[2-4]. This functional distinction of the harmonics allowed us to generate a single-cycle waveform containing the response specific to symmetry, by filtering out the even harmonics in the spectral domain, and then back-transforming the remaining signal, consisting only of odd harmonics, into the time-domain. For our main analysis, we averaged the odd harmonic single-cycle waveforms within a six-electrode region of interest (ROI) over occipital cortex (electrodes 70, 74, 75, 81, 82, 83). These waveforms, averaged over participants, are shown in Figure 2 in the main paper. The same analysis was done for the even harmonics (see Figure S1) and for the odd harmonics within a six electrode ROI over parietal cortex (electrodes 53, 54, 61, 78, 79, 86; see Figure S2). The root-mean square values of these waveforms, for each individual participant, were used to determine whether each of the wallpaper subgroup relationships were preserved in the brain data.

Bayesian Analysis of EEG and Psychophysical data. Bayesian analysis was carried out using R (v3.6.1) (?) with the **brms**

package (v2.9.0 (?) and rStan (v2.19.2 (?)). The data from each experiment were modelled using a Bayesian generalised mixed effect model with wallpaper group being treated as a 16 level factor, and random effects for participant. The EEG data and display thresholds were modelled using log-normal distributions with weakly informative ($N(0, 2)$) priors. After fitting the model to the data, samples were drawn from the posterior distribution for each mean of the EEG response (display duration) for each wallpaper group. These samples were then recombined to calculate the distribution of differences for each pair of subgroup and super-group. These distributions were then summarised by computing the conditional probability that of obtaining a positive (negative) difference, $p(\Delta|data)$.

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331 pair of subgroup and super-group. These distributions were
332 then summarised by computing the conditional probability
333 that of obtaining a positive (negative) difference, $p(\Delta|data)$.
334 For further technical details, please see the supplementary
335 materials where the full R code and model specification can
336 be found.

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338 here, set in a single paragraph. Please do not include any acknowledgments
339 in the Supporting Information, or anywhere else in the
340 manuscript.