

The human visual system preserves the hierarchy of 2-dimensional pattern regularity

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

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

1 **S**ymmetries 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
7 of shapes (? ?), scenes (?) and surface properties (?),
8 as well as the social process of mate selection (?). Most
9 of this work has focused on mirror symmetry or *reflection*,
10 with much less attention being paid to the other fundamental
11 symmetries: *rotation*, *translation* and *glide reflection*. In the
12 two spatial dimensions relevant for images, these four funda-
13 mental symmetries can be combined in 17 distinct ways, the
14 “wallpaper” groups (???). Previous work has focused on
15 four of the wallpaper groups, and used functional MRI to show
16 that rotation symmetries within wallpapers are represented
17 parametrically in several areas in occipital cortex, beginning
18 with visual area V3 (?). This effect is also robust in electroen-
19 cephalography (EEG), whether measured using Steady-State
20 Visual Evoked Potentials (SSVEPs)(?) or event-related
21 paradigms (?). Here we extend on this work by collecting
22 SSVEPs and psychophysical data from human participants
23 viewing the complete set of wallpaper groups. We measure
24 responses in visual cortex to 16 out of the 17 wallpaper groups,
25 with the 17th serving as a control stimulus, with the goal of
26 providing a more complete picture of how wallpaper groups
27 are represented in the human visual system.

28 A wallpaper group is a topologically discrete group of isome-
29 tries of the Euclidean plane, i.e. transformations that preserve
30 distance (?). Wallpaper groups differ in the number and
31 kind of these transformations. In mathematical group theory,
32 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
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 (?). The 17
wallpaper groups thus obey a hierarchy of complexity where
simpler groups are sub-groups of more complex ones (?).

47 The two datasets presented here puts on in the position
48 of being able to assess the extent to which both behavior
49 and brain responses follow that hierarchy. The results show
50 that activity in human visual cortex is remarkably consist-
51 ent with the hierarchical relationships between the wallpaper
52 groups, with SSVEP amplitudes and psychophysical thresh-
53 olds preserving these relationships at a level that is far beyond
54 chance. Visual cortex thus appears to encode all of the fun-
55 damental symmetries using a representational structure that
56 closely approximates the subgroup relationships from group
57 theory. Given that most participants had no knowledge of
58 group theory, the ordered structure of visual responses to
59 wallpaper groups is likely learned implicitly from regularities
in the visual environment.

Significance Statement

Wallpaper groups were discovered in the mid-19th century, and the 17 groups constitute the complete set of possible ways of regularly tiling the 2D-plane. In recent years wallpaper groups have found use in the vision science community, as an ideal stimulus set for studying the perception of symmetries in textures. Here we present brain imaging and psychophysical data on the complete set of wallpaper groups and show the hierarchical organization among wallpaper groups in reflected in both representations in visual cortex and performance on a symmetry detection task. This shows that the visual system is highly sensitive to regularities in textures, and suggest that symmetries may play an important role in texture perception.

PJK and ADFC designed the study, PJK collected EEG data, ADFC collected psychophysical data, PJK and ADFC wrote the paper.

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61 **Results**

62 The stimuli used in our two experiments were multiple ex-
 63 emplar images belonging to each of the wallpaper groups,
 64 generated from random-noise textures, as described in detail
 65 elsewhere (?). Exemplar images from group P1 was used as
 66 control stimuli, and each exemplar from the other 16 groups
 67 had a power-spectrum matched P1 exemplar. The matched P1
 68 exemplars were generated by phase-scrambling the exemplar
 69 images. Because all wallpapers are periodic due to their lat-
 70 tice tiling structure, the phase-scrambled images all belong to
 71 group P1 regardless of group membership of the original exem-
 72 plar. P1 contains no symmetries other than translation, while
 73 all other groups contain translation in combination with one
 74 or more of the other three fundamental symmetries (reflection,
 75 rotation, glide reflection) (?). In our SSVEP experiment,
 76 this stimulus set allowed us to isolate brain activity specific to
 77 the symmetry structure in the exemplar images from activity
 78 associated with modulation of low-level features, by alternat-
 79 ing exemplar images and control exemplars. In this design,
 80 responses to structural features beyond the shared power spec-
 81 trum, including any symmetries other than translation, are
 82 isolated in the odd harmonics of the image update frequency
 83 (? ? ?). Thus, the magnitude of the odd harmonic response
 84 components can be used as a measure of the overall strength
 85 of the visual cortex response.

86 The psychophysical experiment took a distinct but related
 87 approach. In each trial an exemplar image was shown with
 88 its matched control, one image after the other, and the order
 89 varied pseudo-randomly such that in half the trials the original
 90 exemplar was shown first, and in the other half the control
 91 image was shown first. After each trial, participants were told
 92 to indicate whether the first or second image contained more
 93 structure, and the duration of both images was controlled by
 94 a staircase procedure so that a threshold duration could be
 95 computed for each wallpaper group.

96 **Discussion**

97 Here we show that beyond merely responding to the elemen-
 98 tary symmetry operations of reflection (?) and rotation (?),
 99 the visual system explicitly represents hierarchical structure
 100 of the 17 wallpaper groups, and thus the compositions of all
 101 four of the fundamental symmetry transformations (rotation,
 102 reflection, translation, glide reflection) which comprise regular
 103 textures. The RMS measure of SSVEP amplitude, preserves
 104 the complex hierarchy of subgroup relationships among the
 105 wallpaper groups (?). Out of a total of 60 relationships, 53
 106 were preserved in a significant number of participants, and 49
 107 were significant even at a stricter threshold ($p < 0.002$). The
 108 ordering was highly stable in individual participants, with an
 109 average preservation rate of 21 of 25 participants across all 60
 110 relationships (see Figure 3). This remarkable consistency was
 111 specific to the odd harmonics of the stimulus frequency, that
 112 capture the symmetry-specific response (?) and to electrodes
 113 in an ROI over occipital cortex. When the same analysis was
 114 done on the even harmonics of the occipital cortex ROI, the or-
 115 dering of responses was much less apparent (see Figure S2) and
 116 preservation rates much lower (see Figure S4). The odd har-
 117 monics from electrodes in an ROI over parietal cortex, showed
 118 even weaker evidence of preserving the hierarchy among sub-
 119 groups (see Figure S5). Importantly, no relationships were
 120 preserved in either of these control analyses that were not also

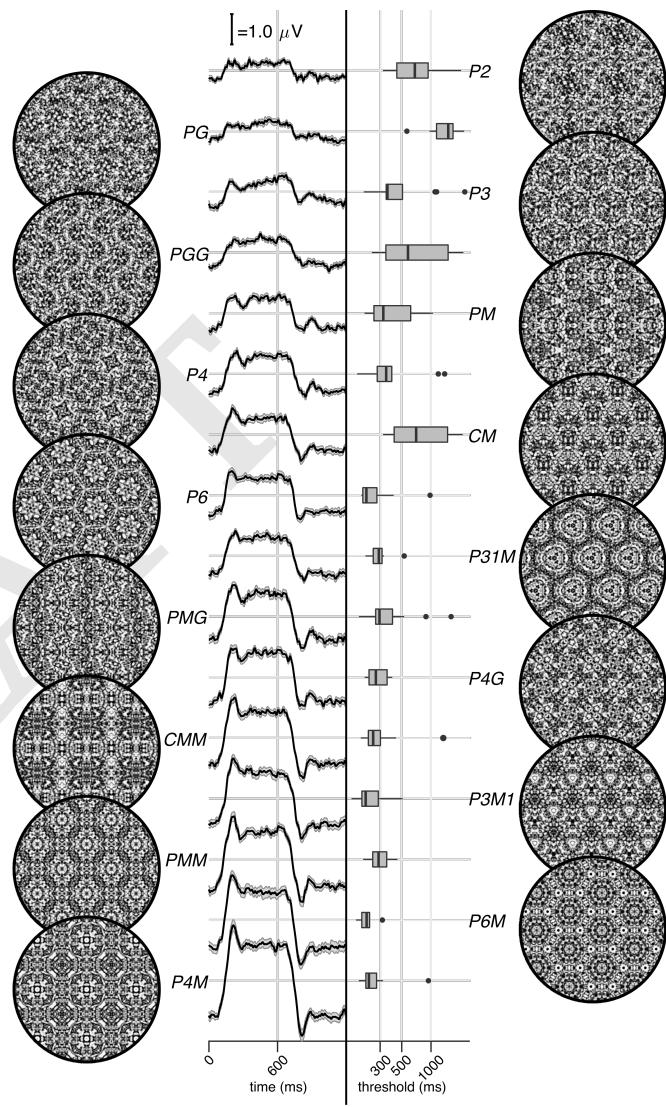


Fig. 1. Posterior distributions for the difference in mean RMS EEG response. Colour coding of the text indicates the index of the subgroup, while the colour of the filled distribution relates to the conditional probability that the difference in means is greater than zero. We can see that 55/64 subgroup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

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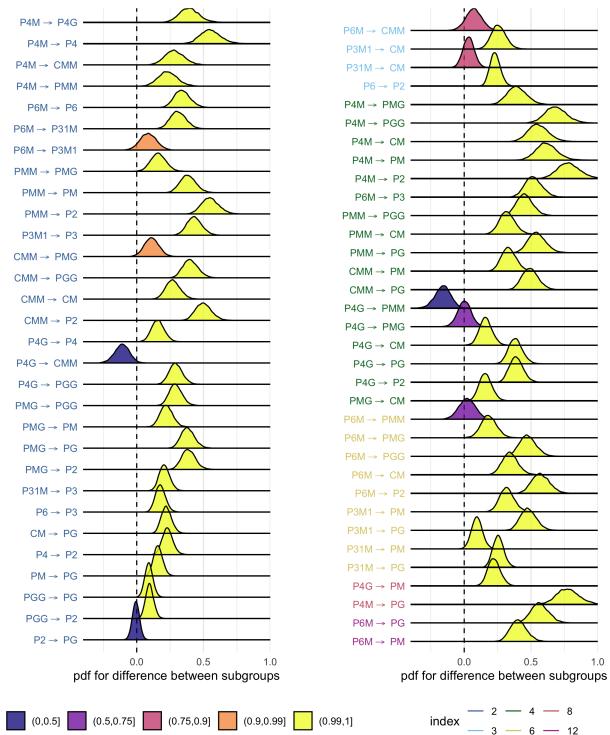


Fig. 2. Posterior distributions for the difference in mean RMS EEG response. Colour coding of the text indicates the index of the subgroup, while the colour of the filled distribution relates to the conditional probability that the difference in means is greater than zero. We can see that 55/64 subgroup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

preserved in the main analysis of the odd harmonics in the occipital cortex ROI. The current data provide a complete description of the visual system's response to symmetries in the 2-D plane. Our design does not allow us to independently measure the response to P1, but because each of the 16 other groups produce non-zero odd harmonic amplitudes (see Figure 2), we can conclude that the relationships between P1 and all other groups, where P1 is the subgroup, are also preserved by the visual system. The subgroup relationships are not obvious perceptually, and most participants had no knowledge of group theory. Thus, the visual system's ability to preserve the subgroup hierarchy does not depend on explicit knowledge of the relationships. Furthermore, behavioral experiments have shown that although naïve observers can distinguish many of the wallpaper groups (?), they are generally error-prone when asked to assign exemplar images to the appropriate group (?). The correspondence between responses in the visual system and group theory that we demonstrate here, may reflect a form of implicit learning that depends on the structure of the natural world. The environment is itself constrained by physical forces underlying pattern formation and these forces are subject to multiple symmetry constraints (?). The ordered structure of responses to wallpaper groups could be driven by a central tenet of neural coding, that of efficiency. If coding is to be efficient, neural resources should be distributed in such a way that the structure of the environment is captured with minimum redundancy considering the visual geometric optics, the capabilities of the subsequent neural coding stages and the behavioral goals of the organism (? ? ? ?). Early work within the efficient coding framework suggested that natural images had a $1/f$ spectrum and that the corresponding redundancy between pixels in natural images could be coded efficiently with a sparse set of oriented filter responses, such as those present in the early visual pathway (? ?). Our results suggest that the principle of efficient coding extends to a much higher level of structural redundancy – that of symmetries in visual images. The 17 wallpaper groups are completely regular, and relatively rare in the visual environment, especially when considering distortions due to perspective and occlusion. Near-regular textures, however abound in the visual world, and can be approximated as deformed versions of the wallpaper groups (?). The correspondence between brain data and group theory demonstrated here may indicate that the visual system represents visual textures using a similar scheme, with the wallpaper groups serving as anchor points in representational space. This framework resembles norm-based encoding strategies that have been proposed for other stimulus classes, most notably faces (?), and leads to the prediction that adaptation to wallpaper patterns should distort perception of near-regular textures, similar to the aftereffects found for faces (?). Field biologist have demonstrated that animals respond more strongly to exaggerated versions of a learned stimulus, referred to as “supernormal” stimuli (?). In the norm-based encoding framework, wallpaper groups can be considered super-textures, exaggerated examples of the near-regular textures that surround us. Artists may consciously or unconsciously create supernormal stimuli, to capture the essence of the subject and evoke strong responses in the audience (?). Wallpaper groups are visually compelling and have been widely used in human artistic expression going back to the Neolithic age (?). If wallpapers are super-textures, their

182 prevalence may be a direct consequence of the strategy the
183 human visual system uses for encoding visual textures.

184 snippets

185 Specifically, the amplitudes of symmetry-specific responses in
186 individual participants ($n = 25$) preserve these relationships
187 at an above-chance level in 88.3% (53 out of 60) of cases.

188 we used a steady-state design, in which exemplar images
189 belonging to 16 of the 17 wallpaper groups alternated with
190 phase-scrambled images of the same group. Exemplars from
191 each of the 16 groups alternated at 0.83 Hz with their corre-
192 sponding set of P1 exemplars, that were matched in terms of
193 their Fourier power spectrum.

194 Thus, the magnitude of the odd harmonic response compo-
195 nents can be used as a distance metric for each group, with
196 distance being measured relative to the simplest group, P1.

197 **Participants.** Twenty-five participants (11 females, mean age
198 28.7 ± 13.3) took part in the EEG experiment. Their informed
199 consent was obtained before the experiment under a proto-
200 col that was approved by the Institutional Review Board of
201 Stanford University. 11 participants (8 females, mean age
202 20.73 ± 1.21) took part in the psychophysics experiment. All
203 participants had normal or corrected-to-normal vision. Their
204 informed consent was obtained before the experiment under a
205 protocol that was approved by the University of Essex's Ethics
206 Committee.

207 **Stimulus Generation.** Exemplars from the different wallpaper
208 groups were generated using a modified version of the method-
209 ology developed by Clarke and colleagues(?) that we have
210 described in detail elsewhere(?). Briefly, exemplar patterns
211 for each group were generated from random-noise textures,
212 which were then repeated and transformed to cover the plane,
213 according to the symmetry axes and geometric lattice spe-
214 cific to each group. The use of noise textures as the starting
215 point for stimulus generation allowed the creation of an al-
216 most infinite number of distinct exemplars of each wallpaper
217 group. For each exemplar image, phase-randomized control
218 exemplars were generated that had the same power spectrum
219 as the exemplar images for each group. The phase scrambling
220 eliminates rotation, reflection and glide-reflection symmetries
221 within each exemplar, but the phase-scrambled images inher-
222 ent the spectral periodicity arising from the periodic tiling.
223 This means that all control exemplars, regardless of which
224 wallpaper group they are derived from, degenerate to another
225 symmetry group, namely P1. P1 is the simplest of the wallpa-
226 per groups, and contains only translations of a region whose
227 shape derives from the lattice. Because the different wallpaper
228 groups have different lattices, P1 controls matched to different
229 groups have different power spectra. Our experimental design
230 takes these differences into account by comparing the neural
231 responses evoked by each wallpaper group to responses evoked
232 by the matched control exemplars.

233 **Stimulus Presentation.** Stimulus Presentation. For the EEG
234 experiment, the stimuli were shown on a 24.5" Sony Trimaster
235 EL PVM-2541 organic light emitting diode (OLED) display
236 at a screen resolution of 1920×1080 pixels, 8-bit color depth
237 and a refresh rate of 60 Hz, viewed at a distance of 70 cm.
238 The mean luminance was 69.93 cd/m² and contrast was 95%²³⁹

239 The diameter of the circular aperture in which the wallpaper
240 pattern appeared was 13.8° of visual angle presented against
241 a mean luminance gray background. Stimulus presentation
242 was controlled using in-house software.

243 For the psychophysics experiment, the stimuli were shown
244 on a 48×27 cm VIEWPixx/3D LCD Display monitor, model
245 VPX-VPX-2005C, resolution 1920×1080 pixels, with a viewing
246 distance of approximately 40cm and linear gamma. Stimulus
247 presentation was controlled using MatLab and Psychtoolbox-3
248 (? ?). The diameter of the circular aperture for the stimuli
249 was 21.5°.

250 **EEG Procedure.** Visual Evoked Potentials were measured using
251 a steady-state design, in which P1 control images alternated
252 with test images from each of the 16 other wallpaper groups[2].
253 Exemplar images were always preceded by their matched P1
254 control image. A single 0.83 Hz stimulus cycle consisted of a
255 control P1 image followed by an exemplar image, each shown
256 for 600 ms. A trial consisted of 10 such cycles (12 sec) over
257 which 10 different exemplar images and matched controls from
258 the same rotation group were presented. For each group type,
259 the individual exemplar images were always shown in the same
260 order within the trials. Participants initiated each trial with
261 a button-press, which allowed them to take breaks between
262 trials. Trials from a single wallpaper group were presented
263 in blocks of four repetitions, which were themselves repeated
264 twice per session, and shown in random order within each
265 session. To control fixation, the participants were instructed
266 to fixate a small white cross in the center of display. To control
267 vigilance, a contrast dimming task was employed. Two times
268 per trial, an image pair was shown at reduced contrast, and the
269 participants were instructed to press a button on a response
270 pad. We adjusted the contrast reduction such that average
271 accuracy for each participant was kept at 85% correct, so that
272 the difficulty of the vigilance task was kept constant.

273 **Psychophysics Procedure.** The experiment consisted of 16
274 blocks, one for each of the wallpaper groups (excluding P1).
275 In each trial, participants were presented with two stimuli
276 (one of which was the wallpaper group for the current block of
277 trials, the other being P1), one after the other (inter stimuli
278 interval of 700ms). After each stimuli had been presented, it
279 was masked with white noise for 300ms. After both stimuli had
280 been presented, participants made a response on the keyboard
281 to indicate whether they thought the first or second contained
282 the most symmetry. Each block started with 10 practise trials,
283 (stimulus display duration of 500ms) to allow participants
284 to familiarise themselves with the current block's wallpaper
285 pattern. If they achieved an accuracy of 9/10 in these trials
286 they progressed to the rest of the block, otherwise they carried
287 out another set of 10 practise trials. This process was repeated
288 until the required accuracy of 9/10 was obtained. The rest of
289 the block consisted of four interleaved staircases (using the
290 QUEST algorithm (?), full details given in the SI) of 30 trials
291 each. On average, a block of trials took around 10 minutes to
292 complete.

293 **EEG Acquisition and Preprocessing.** Electroencephalogram
294 Acquisition and Preprocessing. The time-locked Steady-State
295 Visual Evoked Potentials were collected with 128-sensor HydroCell
296 Sensor Nets (Electrical Geodesics, Eugene, OR) and were band-pass
297 filtered from 0.3 to 50 Hz. Raw data were eval-

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

327 **EEG Analysis.** Response waveforms were generated for each
328 group by selective filtering in the frequency domain. For each
329 participant, the average Fourier coefficients from the two ses-
330 sions were averaged over trials and sessions. The Steady-State
331 Visual Evoked Potentials paradigm we used allowed us to sep-
332 arate symmetry-related responses from non-specific contrast
333 transient responses. Previous work has demonstrated that
334 symmetry-related responses are predominantly found in the
335 odd harmonics of the stimulus frequency, whereas the even
336 harmonics consist mainly of responses unrelated to symme-
337 try, that arise from the contrast change associated with the
338 appearance of the second image[2-4]. This functional distinc-
339 tion of the harmonics allowed us to generate a single-cycle
340 waveform containing the response specific to symmetry, by
341 filtering out the even harmonics in the spectral domain, and
342 then back-transforming the remaining signal, consisting only of
343 odd harmonics, into the time-domain. For our main analysis,
344 we averaged the odd harmonic single-cycle waveforms within
345 a six-electrode region of interest (ROI) over occipital cortex
346 (electrodes 70, 74, 75, 81, 82, 83). These waveforms, averaged
347 over participants, are shown in Figure 2 in the main paper.
348 The same analysis was done for the even harmonics (see Figure
349 S1) and for the odd harmonics within a six electrode ROI over
350 parietal cortex (electrodes 53, 54, 61, 78, 79, 86; see Figure
351 S2). The root-mean square values of these waveforms, for each
352 individual participant, were used to determine whether each
353 of the wallpaper subgroup relationships were preserved in the
354 brain data.

355 **Bayesian Analysis of EEG and Psychophysical data.** Bayesian
356 analysis was carried out using R (v3.6.1) (?) with the `brms`
357 package (v2.9.0) (?) and rStan (v2.19.2 (?)). The data from

each experiment were modelled using a Bayesian generalised
358 mixed effect model with wallpaper group being treated as a
359 16 level factor, and random effects for participant. The EEG
360 data and display thresholds were modelled using log-normal
361 distributions with weakly informative, $\mathcal{N}(0, 2)$, priors. After
362 fitting the model to the data, samples were drawn from the pos-
363 terior distribution for each mean of the EEG response (display
364 duration) for each wallpaper group. These samples were then
365 recombined to calculate the distribution of differences for each
366 pair of subgroup and super-group. These distributions were
367 then summarised by computing the conditional probability of
368 obtaining a positive (negative) difference, $p(\Delta|data)$.
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370 For further technical details, please see the supplementary
371 materials where the full R code, model specification, prior and
372 posterior predictive checks, and model diagnostics, can be
373 found.

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