

# <sup>1</sup> Perceptual Similarities Among Wallpaper Group Exemplars

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## <sup>7</sup> Abstract

<sup>8</sup> Symmetries are abundant within the visual environment, and many animals species are  
<sup>9</sup> sensitive to visual symmetries. Wallpaper groups a class of 17 regular textures that each contain  
<sup>10</sup> a distinct combination of the four fundamental symmetries, translation, reflection, rotation  
<sup>11</sup> and glide reflection, and together represent the complete set of possible symmetries in two-  
<sup>12</sup> dimensional images. Wallpapers are visually compelling and elicit responses in visual brain  
<sup>13</sup> areas that precisely capture the symmetry content of each group, in humans and other primates.  
<sup>14</sup> Here we ask to what extent exemplars from the same wallpaper group are perceptually similar.  
<sup>15</sup> We algorithmically produce a set of well-matched exemplars from 5 of the 17 wallpaper groups  
<sup>16</sup> and instructed participants to freely sort the exemplars from each group into as many subsets  
<sup>17</sup> as they wished based on any criteria they saw appropriate.  $P_1$ , the simplest of the 17 groups,  
<sup>18</sup> was consistently rated more self-similar than any other group, while the other four groups,  
<sup>19</sup> although varying in symmetry content, were comparable in self-similarity. Our results suggest  
<sup>20</sup> that except for the most extreme case ( $P_1$ ), self-similarity of wallpaper groups is not directly  
<sup>21</sup> tied to symmetry content.

## <sup>22</sup> Introduction

<sup>23</sup> Symmetry has been recognized as important for human visual perception since the late 19th cen-  
<sup>24</sup> tury (Mach, 1959). In the two spatial dimensions relevant for images, symmetries can be combined  
<sup>25</sup> in 17 distinct ways, *the wallpaper groups* (Fedorov, 1891; Polya, 1924; Liu et al., 2010). Wallpaper  
<sup>26</sup> groups are different from the stimuli typically used to probe the role of symmetry in visual per-  
<sup>27</sup> ception in two ways: First, they contain combinations of the four fundamental symmetry types  
<sup>28</sup> translation, reflection, rotation and glide reflection, rather than just reflection or mirror symmetry,  
<sup>29</sup> which has been the focus of most vision research. Second, the symmetries in wallpaper groups are  
<sup>30</sup> repeated to tile the plane, rather than positioned at a single image location as is usually the case.  
<sup>31</sup> These differences, and the fact that wallpaper groups together form the complete set of symme-  
<sup>32</sup> tries possible in the two-dimensional image plane, make wallpapers an interesting stimulus set for  
<sup>33</sup> studying perception of visual symmetries.

34 Brain imaging studies using functional MRI (Kohler  
 35 et al., 2016) and EEG (Kohler et al., 2018; Kohler and  
 36 Clarke, 2021) has shown that the human visual system  
 37 carries detailed and precise representations of the sym-  
 38 metries within the individual wallpaper groups, and  
 39 functional MRI evidence from macaque monkeys re-  
 40 veal similar representations in analogous areas of the  
 41 macaque visual system (Audurier et al., 2021).

42 These representations, complex as they are, do not  
 43 appear to be readily available for driving conscious  
 44 behaviour: Humans have limited intuitive sense of  
 45 group membership for wallpaper group exemplars, as  
 46 evidenced by behavioral experiments showing that al-  
 47 though naïve observers can distinguish many of the wall-  
 48 paper groups (Landwehr, 2009), they tend to sort exem-  
 49 plars into fewer (4-12) sets than the number of wallpaper  
 50 groups, often placing exemplars from different wallpaper  
 51 groups in the same set (Clarke et al., 2011). Wallpaper  
 52 groups are nonetheless visually compelling, and anec-  
 53 dotally we have observed that exemplars from a given  
 54 group can be quite perceptually diverse. This obser-  
 55 vation inspired the current study, in which we use the  
 56 behavioral sorting approach to probe the perceptual self-  
 57 similarity of different exemplars from the same wallpa-  
 58 per group, and assess the extent to which self-similarity  
 59 varies across five groups.

60 We algorithmically generated 20 well-matched exem-  
 61 plars from each group (see Figures 1 and 2 for a selection  
 62 of the exemplars, and the **Materials and Methods** sec-  
 63 tion for details on how they were generated) and printed  
 64 them out on white cardstock. We then gave participants  
 65 the 20 cards with exemplars from each wallpaper group,  
 66 and asked them to freely sort them into as many sub-  
 67 sets as they wished based on any criteria they saw ap-  
 68 propiate. This approach allowed us to compare the five  
 69 wallpaper groups, both in terms of how many subsets  
 70 participants generated, and also in terms of the *Jaccard*  
 71 index, a summary statistic capturing the similarity across  
 72 exemplar pairs for each group. Within each group, we  
 73 were also able to identify exemplar pairs that were rated  
 74 as highly similar and highly dissimilar. Our main con-

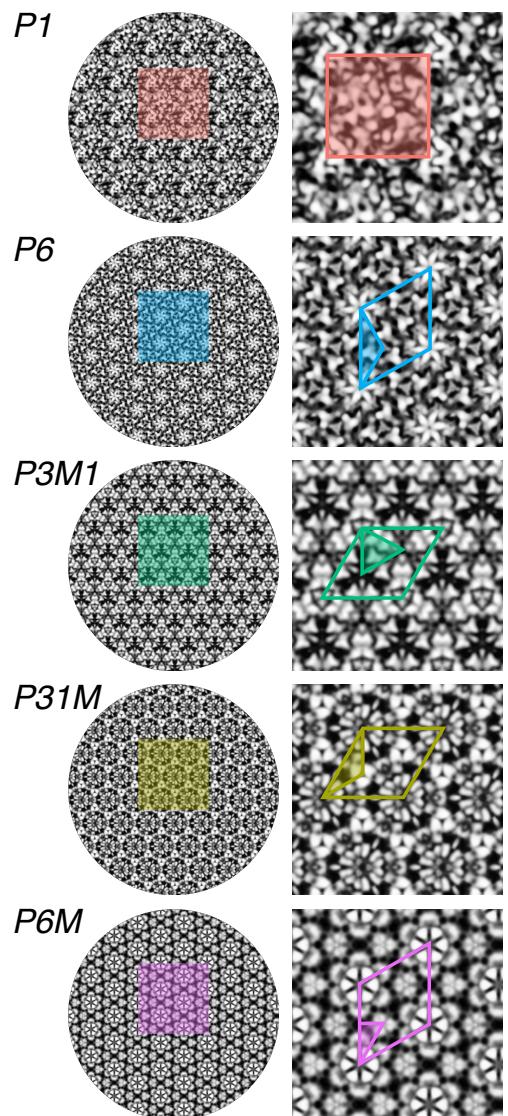


Figure 1: The fundamental region and lat-  
 tice structure of the five wallpaper groups  
 used in the study. The complete wallpa-  
 per is shown in the left-hand column with  
 a shaded region that is repeated and en-  
 larged in the right-hand column. The col-  
 ored outline in the enlarged region indi-  
 cates the repeating lattice for each group,  
 while the shaded area indicates the funda-  
 mental region (see text). For  $P_1$  the funda-  
 mental region covers the entire lattice.  
 Note that even though  $P_6$  and  $P_{31}M$  have  
 the same fundamental region and lattice  
 shapes, they differ in terms of the symme-  
 tries present within the lattice - most no-  
 tably,  $P_{31}M$  contains reflection symmetry  
 while  $P_6$  does not. The symmetry content  
 of each group is detailed on the wallpaper  
 group wikipedia page.

clusion is that  $P_1$  was systematically less self-similar than the any other groups, while the other four other groups could not be distinguished on these measures. We also show that for all five groups, participants consistently group certain pairs of exemplars together, although the number of consistent pairs varies among groups. Our results open the door to further investigations into the factors that drive perceptual similarity among wallpaper groups exemplars, and indeed among exemplars from different classes of structured patterns.

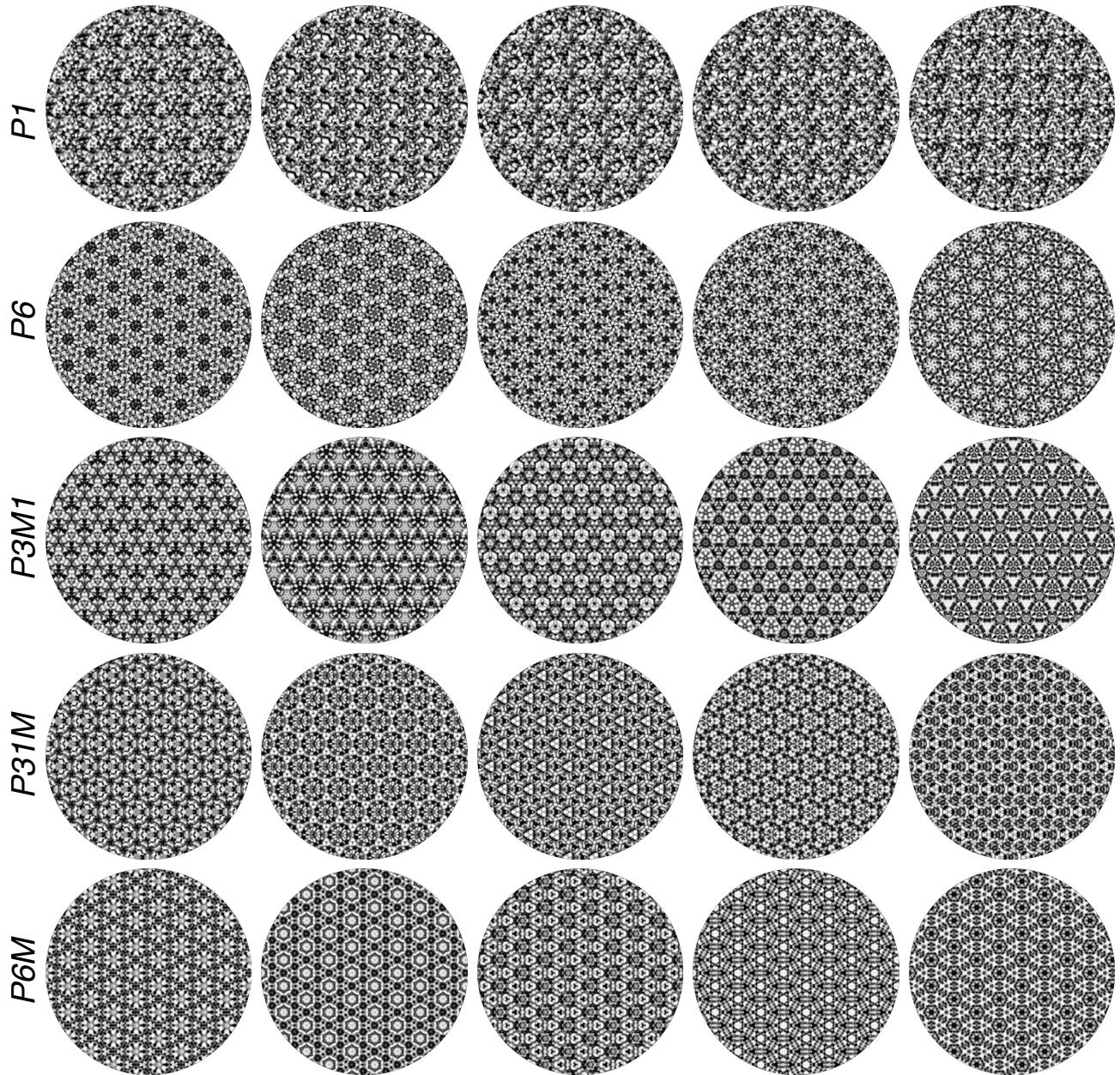


Figure 2: 5 of the 20 exemplars used for each group are shown to highlight the diversity among exemplars.

## 81 Results

Wallpaper group  $P_1$  was less self-similar than the other four groups. This was evident in the number of sets generated for this group across participants, which was lower for  $P_1$  (median = 3) than for the other groups (median = 4-5, see Figure 3). We confirmed this observation statistically by running a repeated measures analysis of variance (ANOVA) with group as a fixed factor and participant as a random factor, which revealed a significant effect of group ( $F(4,124) =$

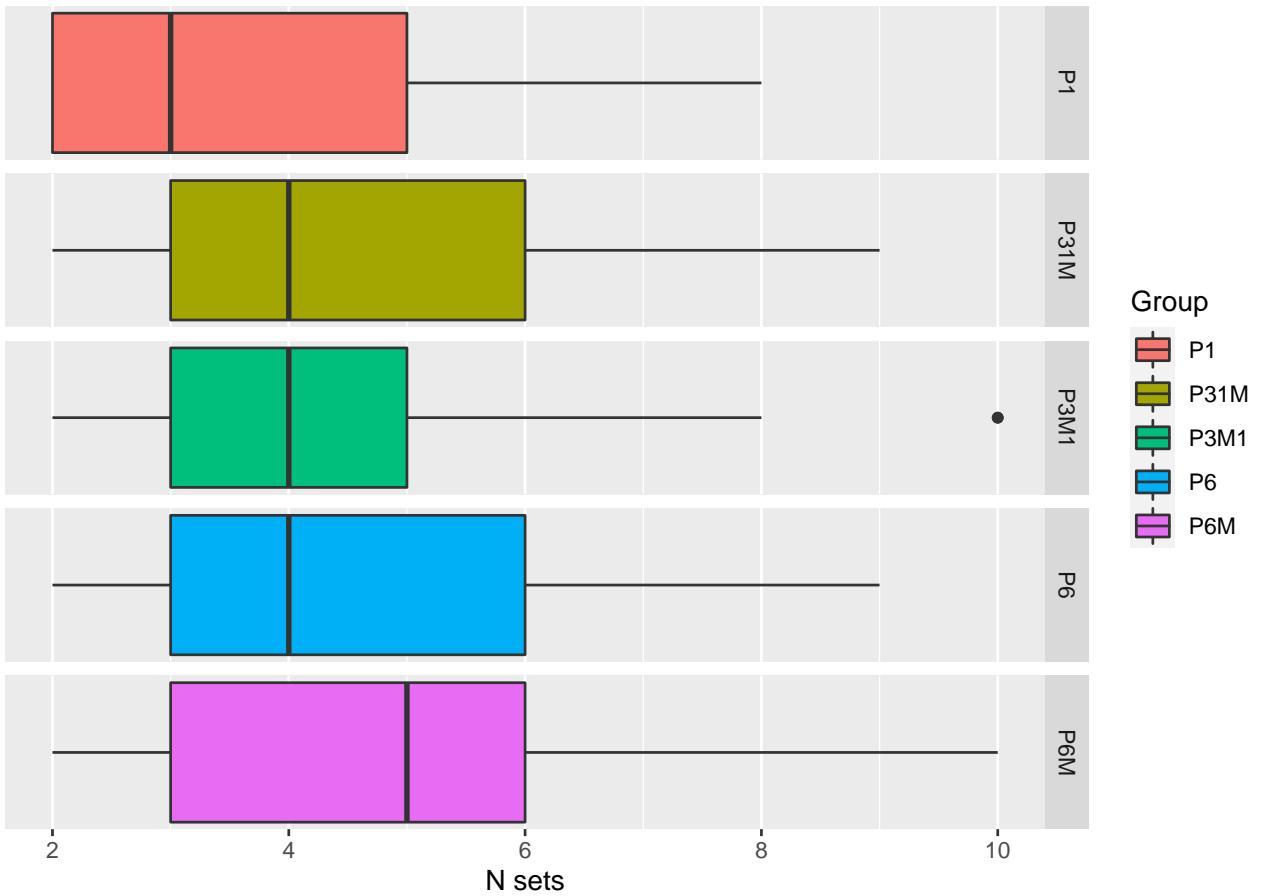


Figure 3: Boxplots showing the number of subsets generated by participants for each of the wallpaper groups. The lower box boundary is the 25th percentile. The dark line in the box is the median. The upper box boundary is the 75th percentile. The “whiskers” show -/+ the interquartile range \* 1.5.

7.330,  $p < 0.0001$ ). Post-hoc pairwise  $t$ -tests showed that the mean number of sets was lower for  
 88  $P_1$  than all other groups, but no other means differed. Next, we computed the Jaccard index (see  
 89 **Materials and Methods**) across participants for every pairwise combination of exemplars in each  
 90 group. This provides a measure of the similarity between exemplars within each group.  $P_1$  had  
 91 systematically higher Jaccard indices than the four other groups (see Figure 4), as confirmed by an  
 92 ANOVA with wallpaper group as a factor. The analysis revealed a statistically significant effect of  
 93 group ( $F(4, 495) = 20.178, p < 0.0001$ ). Post-hoc pairwise  $t$ -tests showed that  $P_1$  had higher Jaccard  
 94 indices than all other groups ( $p < 0.0001$ ). The fact that the group ( $P_1$ ) for which fewer subsets  
 95 were generated also had higher Jaccard indices than the other groups illustrates the inherent link  
 96 between the two measures: For wallpaper groups where the 20 exemplars are sorted into fewer  
 97 subsets, each individual exemplar pair are more likely to be members of the same subset, and  
 98 less likely to be members of distinct subsets, which in turn leads to higher Jaccard indices. Our  
 99 pairwise  $t$ -tests also showed that  $P_{31M}$  had lower Jaccard indices than  $P_6$  ( $p = 0.037$ ). This effect  
 100 is relatively weak, but may reflect real differences in how consistently exemplars were grouped  
 101 together across participants. We will explore this idea more in depth shortly, but for now we can  
 102 conclude that out of the five groups tested,  $P_1$  is the only one that can be reliably differentiated  
 103 based on our measures, being higher on self-similarity among the exemplars, and thus lower on  
 104 diversity among exemplars.

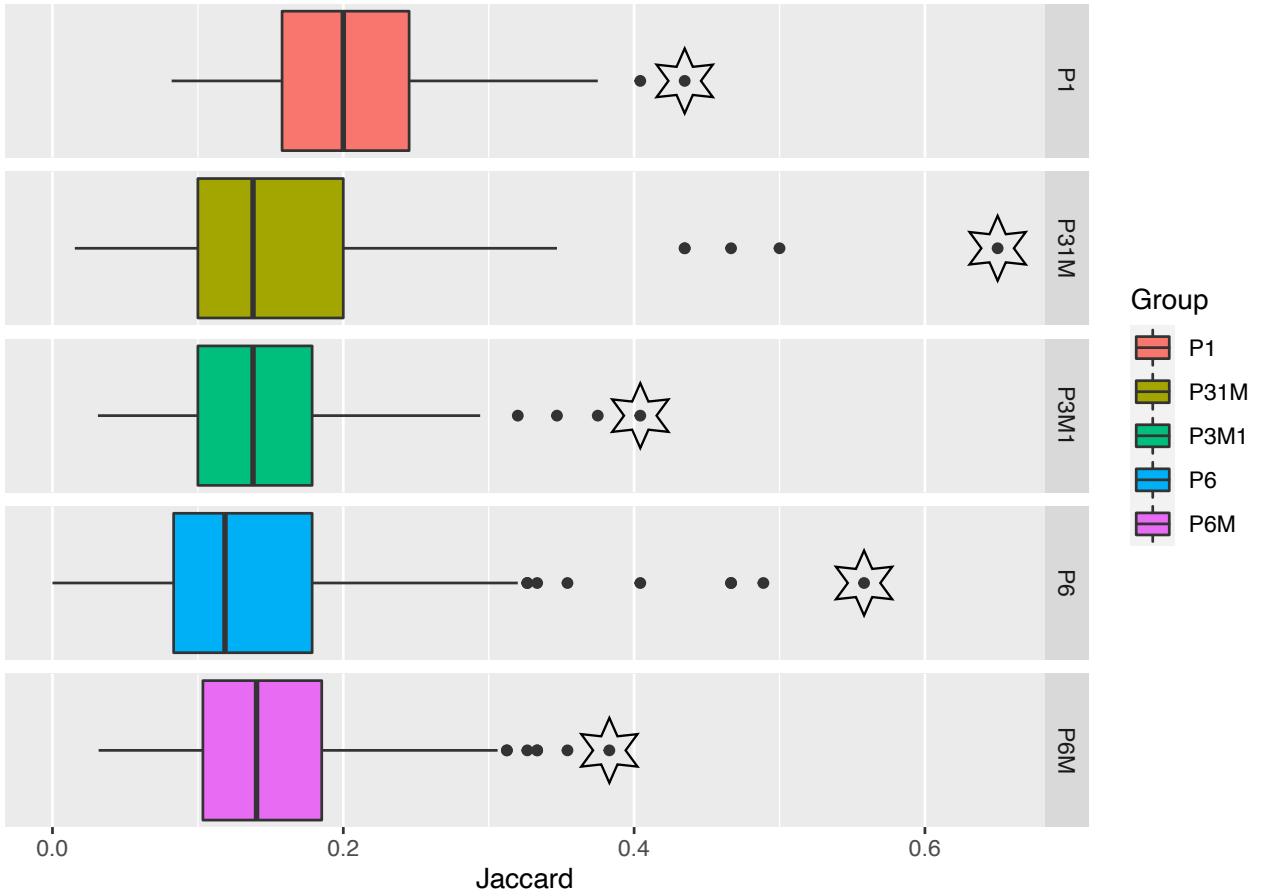


Figure 4: Boxplots showing Jaccard indices for every pairwise combination of exemplars in each of the wallpaper groups. Note that each data point here is the Jaccard index for a particular exemplar pair calculated across participants, unlike Figure 3 where each data point is a participant. The box boundary and whiskers follow the same logic as in Figure 3. The exemplar pairs with the highest Jaccard indices have been highlighted with stars. Those outlier pairs are explored further in Figure 6.

In order to quantify the extent to which exemplars were consistently grouped together, we ran a permutation analysis in which exemplar labels were shuffled among the sets generated for each participant (see Materials and Methods). This provides, for each group, the expected distribution of Jaccard indices for every pairwise combination of exemplars, if exemplars were assigned randomly to subsets, and allows us to compute an empirical  $z$ -score that expressed the extent to which a given pair of exemplars deviates from random assignment. Because the random distribution is generated by shuffling exemplars across the specific sets generated by each participant for each group, this  $z$ -score is independent of the number of sets. If for a given group, none of the pairs deviate significantly from the random distribution, it would indicate that no exemplar pairs were consistently grouped together across participants. To estimate the extent to which this is the case, we look at the distribution of  $z$ -scores across the pairs for each group, as plotted below.

group	consistent pairings	
	$p < 0.01$	$p < 0.0001$
$P_1$	6	1
$P_{31M}$	17	10
$P_{3M1}$	12	3
$P_6$	17	11
$P_{6M}$	15	4

Table 1: Number of consistent pairings at two different  $\alpha$ -levels for the five groups.

ted in Figure 5, and count the number of pairs for each group for which the  $p$ -value associated with the threshold exceeds a given  $\alpha$  value. At a threshold of  $\alpha = 0.01$ , several pairs survive for all groups, and even at a much more conservative criterion of  $\alpha = 0.0001$  most groups have more than one pairing that survives (see Table 1). It is worth noting that the latter threshold ( $\alpha = 0.0001$ ) is lower than the  $\alpha$  associated with a Bonferroni correction within group, given that there are 190 pairs per group:

$$\alpha = \frac{0.05}{190} = 0.0000263$$

So we conclude that for several exemplar pairs participants are consistent in how they tend to pair the exemplars. It is interesting to consider that this measure of consistency might provide another way of differentiating wallpaper groups in terms of perceptual self-similarity. While groups  $P_{31}M$ ,  $P_3M_1$ ,  $P_6$  and  $P_6M$  have comparable Jaccard scores (see Figure 4), they differ in the number of consistent pairings, with  $P_{31}M$  and  $P_6$  producing more consistent pairs than the other two (see Figure 5).

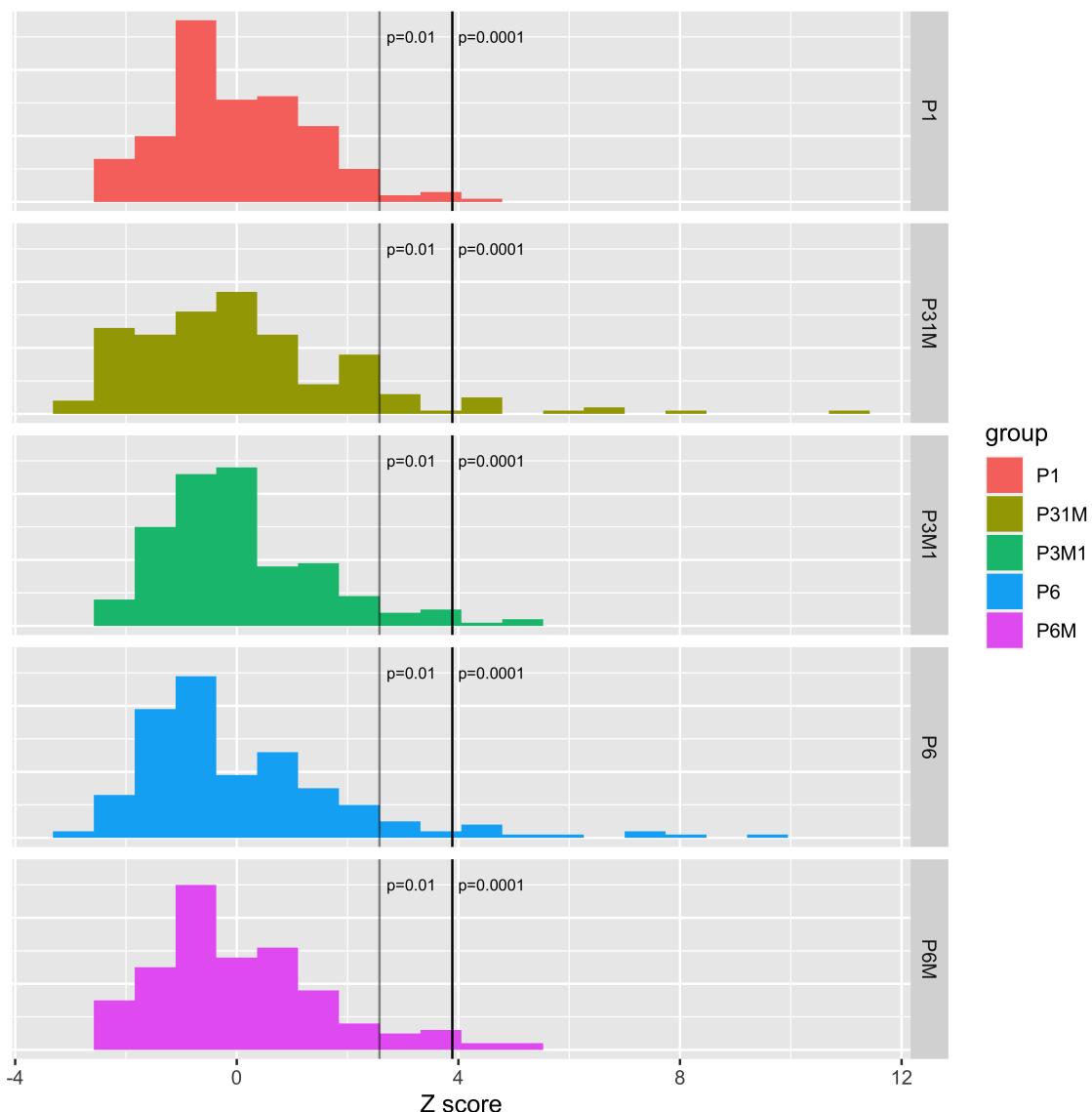


Figure 5: Distribution of  $z$ -scores across the 190 pairs in each of the five wallpaper groups. The two lines indicate the  $z$ -scores associated with  $\alpha$  of 0.01 and 0.0001, respectively.

The Jaccard indices also allow us to focus on exemplar pairs that have a high level of similarity

relative to the rest of the pairs in the set. We do this by identifying outliers pairs from each group in term of Jaccard indices, as identified with stars in Figure 4. Because the Jaccard indices are computed across participants, these outliers are also among the pairs most consistently sorted together, as identified in Figure 5. For each exemplar in each outlier pair, we can visualize the pairwise similarity (as measured by the Jaccard index) to every other exemplar in the set (see Figure 6). That is, we can visualize portions of the network of perceived similarity within a set of exemplars. Future work could probe the extent to which networks of perceived similarity have similar structure across wallpaper groups and examine what perceptual features best account for participants' perceptions of exemplar similarity.

## Discussion

Previous work has demonstrated that visual cortex of both humans and macaque monkeys carries highly detailed representations of the symmetries within wallpaper groups, as evidenced by systematic differences in the magnitude of the response elicited by different groups (Kohler et al., 2016; Kohler and Clarke, 2021; Audurier et al., 2021). This distinction between groups that can also be observed in psychophysical threshold measurements (Kohler and Clarke, 2021), although observers may not have a strong awareness of the wallpaper group membership of individual exemplars (Clarke et al., 2011). In the current study, we explored a new piece of the story of how wallpaper groups are processed by the visual system, namely the issue of how self-similar different exemplars from the *same* wallpaper group appear to untrained observers. We tested this by asking participants to spontaneously sort 20 exemplars from each of five wallpaper groups into different subsets.

Our first finding concern the number of subsets generated for each group. We find that  $P_1$  is divided into fewer subsets than the other four groups. This indicates that the limited complexity of this group, which contains only translation symmetry, has a direct effect of the number of distinct subsets. The relationship between complexity / symmetry content and number of subsets produced is not straightforward, however, as indicated by the fact that  $P6M$  is not consistently grouped into more subsets than  $P6$ ,  $P3M_1$  and  $P3_1M$ , despite the fact that these other groups all contain fewer symmetries than  $P6M$  - and generate weaker brain activity (Kohler and Clarke, 2021). We can speculate that this lack of further differentiation is a result of an upper limit on how additional complexity can influence perceptual self-similarity.

We also compute Jaccard indices that for every possible exemplar pair expresses the frequency of those two exemplars being grouped together. As described above, the average Jaccard index for a group is inherently linked to the number of subsets produced for that group, because fewer subsets mean that exemplars are more likely to be made members of the same pair, and less likely to be made members of the same pair. It is therefore not surprising that we find the same general pattern for Jaccard indices and number of subsets, namely that  $P_1$  has higher indices than the other groups. The advantage of the Jaccard indices, however, is that they allow us to conduct a permutation analysis that quantifies the extent to which pairs of exemplars are consistently grouped together across participants, independent of the number of sets produced for

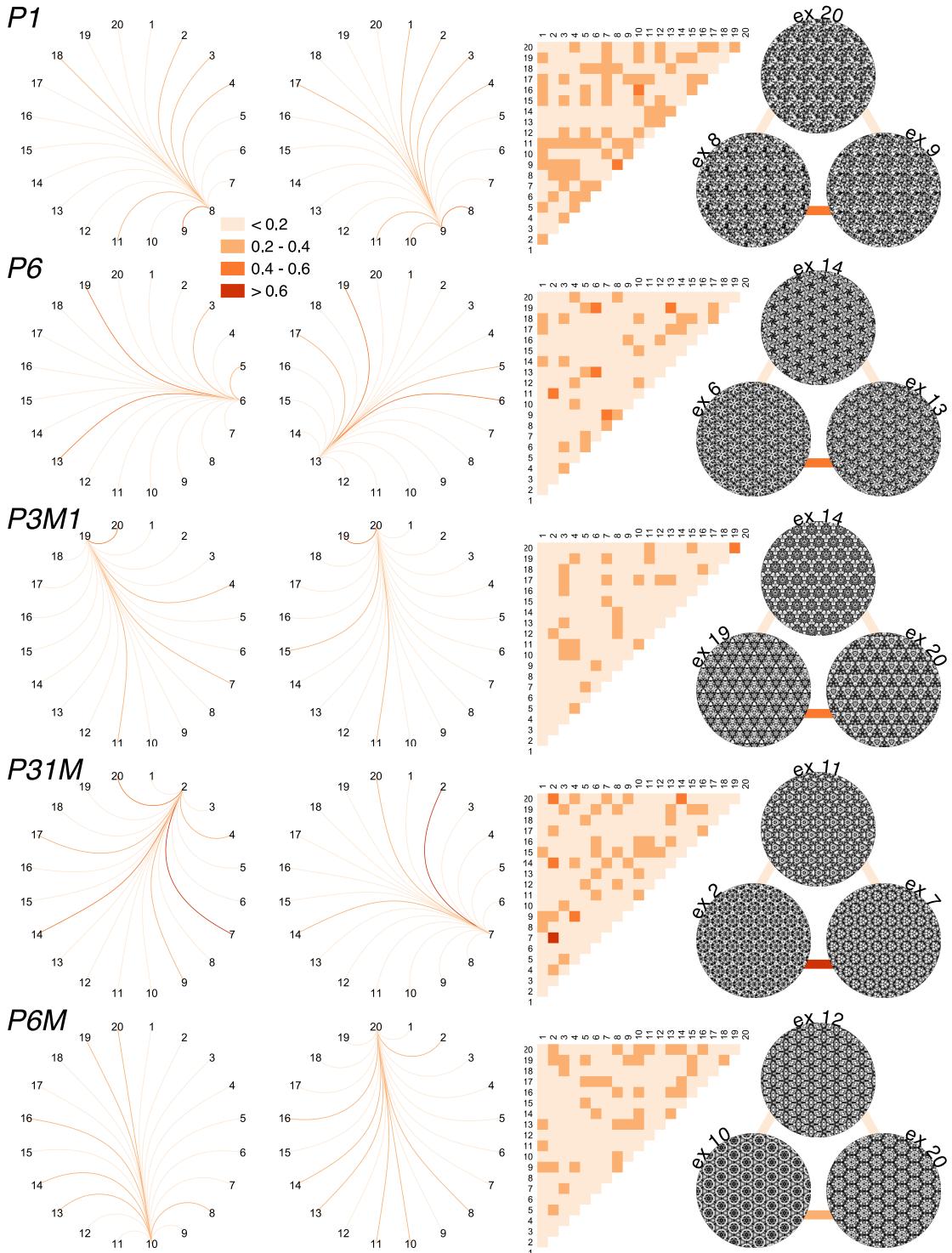


Figure 6: For each wallpaper group, we identified the two most self-similar exemplars, the same pair that is indicated by the right-most datapoint for each group in Figure 5. The two circular network plots are showing the pairwise similarities between those two exemplars and every other exemplar in the set. The pairwise similarities across all exemplars are plotted as a similarity matrix and on the rightmost side of the plot, the two most self-similar exemplars (bottom) are plotted with the exemplar that was least similar to both (top). The connecting lines between the exemplars indicate the similarity.

174 a given group. It is important to note that consistency in the choice of which exemplars to group  
 175 into subsets is not an unavoidable consequence of our experimental design, and does not follow  
 176 naturally from the results described so far. It would be perfectly possible for participants to group  
 177 the sets together, producing fewer subsets for  $P_1$  as observed, but exhibit no consistency across

178 participants at all. That is not what we see, however. Even when setting a conservative threshold,  
179 all five groups produce one or more pairs that are consistently grouped together, demonstrating  
180 that the sorting of exemplars into subsets is not done randomly or arbitrary, and that different  
181 individuals at least to some extent agree on which exemplars belong together. Because our  
182 measure of consistency is independent of the number of subsets produced for a given group, it  
183 allows us to show that although  $P_1$  has the highest overall Jaccard indices (as a result of the fewer  
184 sets produced for this group), it in fact produces fewer consistent pairs than other groups (see  
185 Table 1).

## 186 Materials and Methods

### 187 Participants

188 33 participants (9 Male, 24 Female), ranging in age between 18 and 35 completed this study. All  
189 participants had self reported 20/20 or corrected to 20/20 vision. We obtained written consent to  
190 participate from all participants under procedures approved by the Institutional Review Board  
191 of The Pennsylvania State University (#38536). The research was conducted according to the  
192 principles expressed in the Declaration of Helsinki.

### 193 Stimulus Generation

194 Five wallpaper groups ( $P_1$ ,  $P_3M_1$ ,  $P_{31}M$ ,  $P_6$  and  $P_6M$ ) that has previously been shown to be high in  
195 self-similarity (Clarke et al., 2011), were selected. 20 exemplars from each of these five wallpaper  
196 groups were generated using a modified version of the methodology developed by Clarke and  
197 colleagues (Clarke et al., 2011) that we have described in detail elsewhere (Kohler et al., 2016).  
198 Briefly, exemplar patterns for each group were generated from random-noise textures, which  
199 were then repeated and transformed to cover the plane, according to the symmetry axes and  
200 geometric lattice specific to each group. The use of noise textures as the starting point for  
201 stimulus generation allowed the creation of an almost infinite number of distinct exemplars of  
202 each wallpaper group. To make individual exemplars as similar as possible we replaced the power  
203 spectrum of each exemplar with the median across exemplars within a group. These images  
204 were printed onto white cardstock and cut into squares, allowing participants to manipulate the  
205 orientation of the images during the sorting tasks. Five exemplars from each group are shown  
206 (in reduced size) in Figure 2.

### 207 Procedure

208 Participants were presented with the 20 exemplars of a single wallpaper group (i.e.  $P_1$ ,  $P_3M_1$ ,  
209  $P_{31}M$ ,  $P_6$ ,  $P_6M$ ) and instructed to sort them into subsets by placing them into piles. Participants  
210 were advised to sort the exemplars into as many piles as they deemed necessary based on  
211 whatever criteria they desired. There were no time constraints placed on this sorting task, and  
212 the participants were allowed to move exemplars between piles until they were satisfied with  
213 their classification. This method was then repeated for the remaining four wallpaper groups for

214 each participant, with group presentation order randomized between participants. These tasks  
215 were carried out on a large table with sufficient space to randomly lay out all twenty exemplars  
216 of each set, illuminated by normal overhead room lighting. Upon completion of each sorting  
217 task, participants were asked to verbalize which features they used to sort the exemplars. After  
218 completion of all five sorting tasks, participants were asked which if they had a distinct method  
219 for sorting the images, and if any wallpaper group was particular easy or difficult to sort.

## 220 Generating the Jaccard Index

221 The data was prepared for analysis by creating one binary variable for each subset created by  
222 each participant within a sorting task. Then, each exemplar was assigned a value of one (1) if  
223 it was included in a subset, or a value zero (0) if it was not. Next, the similarity of each pair of  
224 exemplars within a sorting task was calculated using the Jaccard index, a measure of similarity  
225 and diversity for binary data. This index is calculated by the equation

$$J = \frac{x}{x + y + z}$$

226 with x representing the number of subsets that contained both exemplars, and y and z the number  
227 of subsets that contain only one exemplar of the pair (Capra, 2005), across participants. Thus,  
228 the Jaccard index is the ratio of the number of subsets containing both exemplars of a pair to the  
229 number of subsets containing at least one of the exemplars of a pair, thereby excluding subsets  
230 with joint absences.

## 231 Permutation Analysis

232 The permutation analysis involved generating a randomized dataset, as follows. For each partic-  
233 ipant and wallpaper group, we randomized which specific exemplars were sorted together. This  
234 retained the basic structure of each participants' sorting data—the number of subsets created—  
235 but randomized the relationship between specific wallpaper exemplars that were sorted together  
236 across the participants. We then created 1,000 such permuted datasets, and calculated the Jac-  
237 card index for each exemplar pair within each group for each of the permuted datasets. This  
238 permitted the calculation of an *empirical* Jaccard index based on the permuted data from which  
239 distributional statistics like z could be calculated. The *observed* Jaccard indices for each exemplar  
240 pair were then compared to the empirically-derived reference distribution to determine which  
241 exemplar pairs were sorted together more frequently than chance would predict.

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