

# Effects of background and contour luminance on the hue and brightness of the Watercolor effect

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3	brightness of the water color effect
4 5 6	Peggy Gerardin <sup>2,</sup> , Michel Dojat <sup>3</sup> , Kenneth Knoblauch <sup>2,</sup> , Frédéric Devinck <sup>1*</sup> ,
7	
8	<sup>1</sup> Université Rennes 2, LP3C, EA 1285, 35000 Rennes, France
9	<sup>2</sup> Univ Lyon, Université Claude Bernard Lyon 1, Inserm, Stem Cell and Brain
10	Research Institute U1208, 69500 Bron, France. 3 Université Lyon 1, 69003
11	Lyon, France
12	<sup>3</sup> Univ. Grenoble Alpes, Inserm, CHU Grenoble Alpes, GIN, 38000 Grenoble,
13	France
14	
15	* Corresponding author. Tel.: +33-2-99-19-59; Fax: +33-2-99-14-19-54
16	e-mail address: frederic.devinck@univ-rennes2.fr
17	Present address: Département de Psychologie, Université Rennes 2, 35043
18	Rennes Cedex, France.
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#### 21 Abstract

22

Conjoint measurement was used to investigate the joint influences of the 23 luminance of the background and the inner contour on hue- and brightness filling-24 in for a stimulus configuration generating a water-color effect (WCE), i.e., a 25 wiggly bi-chromatic contour enclosing a region with the lower luminance 26 component on the exterior. Two stimuli with the background and inner contour 27 luminances covarying independently were successively presented, and in separate 28 experiments, the observer judged which member of the pair's interior regions 29 contained a stronger hue or was brighter. Braided-contour control stimuli that 30 generated little or no perceptual filling-in were also used to assess whether 31 observers were judging the interior regions and not the contours themselves. 32 Three nested models of the contributions of the background and inner contour to 33 the judgments were fit to the data by maximum likelihood and evaluated by 34 35 likelihood ratio tests. Both stimulus components contributed to both the hue and brightness of the interior region with increasing luminance of the inner contour 36 generating an assimilative filling-in for the hue judgments but a contrast effect for 37 the brightness judgments. Control analyses showed negligible effects for the 38 order of the luminance of the background or inner contour on the judgments. An 39 40 additive contribution of both components was rejected in favor of a saturated model in which the responses depended on the levels of both stimulus 41 components. For the hue judgments, increased background luminance led to 42

greater hue filling-in at higher luminances of the interior contour. For the brightness judgments, the higher background luminance generated less brightness filling-in at higher luminances of the interior contour. The results indicate different effects of the inner contour and background on the induction of the brightness and coloration percepts of the WCE, suggesting that they are mediated by different mechanisms.

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51 <u>Keywords</u>: filling-in, color assimilation, watercolor effect, color appearance,

52 scaling, conjoint measurement, MLCM.

## 53 1. Introduction.

Color appearance is not determined only by the local light signals from each 54 object but is also influenced by global contextual features. The watercolor 55 effect (WCE) is an interesting phenomenon for studying such processes (Pinna, 56 1987; Pinna et al., 2001). A pair of wiggly contours composed of a light 57 chromatic contour (e.g., orange) surrounded by a darker chromatic contour (e.g., 58 purple) bounding an achromatic surface area elicits a filling-in of the hue of the 59 lighter contour over the entire enclosed area (Figure 1a). The WCE is 60 distinguished from other assimilation illusions by its large spatial extent; the 61 phenomenon has been observed over distances of up to 45 deg (Pinna et al., 62 2001). In addition to the assimilative color spreading, the subjectively colored 63 area is perceived as figure while the surrounding area appears as ground (Pinna 64 et al., 2003; Pinna & Tanca, 2008; Tanca & Pinna, 2008). 65

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- 67

#### Figure 1 about here

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69 Studies of the WCE have typically examined the effects of the inducer 70 configuration producing the WCE. For example, the intensity of the filling-in 71 percept appears greater with increases in luminance contrast between the inner 72 and outer contours for an achromatic WCE (Cao et al., 2011) and for a WCE 73 that has both luminance and chromatic components (Devinck et al., 2005; 74 Devinck & Knoblauch, 2012). Devinck et al. (2005) noted that observers did

not need to modify significantly the luminance of the enclosed area in a 75 matching experiment. Other critical characteristics of the inducing contours that 76 modulate the strength of the WCE include the continuity and contiguity of the 77 contour pairs (Devinck & Spillmann, 2009; Devinck & Knoblauch, 2012). 78 Recent demonstrations of the sensitivity of the phenomenon to contour 79 adaptation provide additional support for a role of contour integration 80 mechanisms in the WCE (Coia & Crognale, 2017). The strength of the 81 phenomenon was found to be size-tuned with the strongest WCE observed for a 82 contour width of about 15 arcmin and was optimal for equal contour widths 83 (Devinck et al., 2014a). While the WCE has been reported for linear contours 84 (Pinna et al., 2001), its strength is nearly independent of the amplitude of 85 contour undulation but increases with contour frequency up to an asymptotic 86 level (Gerardin et al., 2014). Finally, Pinna et al. (2001) demonstrated that 87 several different color pairs can generate the coloration effect (see also Devinck 88 Specifically, Devinck et al. (2006) demonstrated that the et al., 2005). 89 coloration effect is stronger when the chromatic contrast is larger. Thus, the 90 coloration effect depends on a conjunction of chromatic and luminance contrasts 91 92 but also on spatial parameters of the inner and outer contours.

The WCE is perceptually salient but has proved difficult to quantify with precision showing large variability within and across observers (Cao et al., 2011; Devinck et al., 2005; von der Heydt & Pierson, 2006). More recently, the WCE was quantified by using paired-comparison methods that have been

extended to estimate perceptual scales within a signal detection framework 97 (Devinck & Knoblauch, 2012). Two such procedures are Maximum Likelihood 98 Difference Scaling or MLDS (Maloney & Yang, 2003; Knoblauch & Maloney, 99 2008, 2012) and Maximum Likelihood Conjoint Measurement or MLCM (Ho et 100 al., 2008; Knoblauch & Maloney, 2012). Difference scaling is useful for 101 measuring perceptual strength along a single physical dimension, whereas 102 conjoint measurement was conceived to assess the combined effects of several 103 dimensions on appearance (Falmagne, 1985; Knoblauch & Maloney, 2012; 104 Krantz et al., 1971; Luce & Tukey, 1964; Roberts, 1979). MLCM has been 105 successfully applied to estimate perceptual scales associated with different sets 106 of physical continua including surface material properties (Ho, Landy & 107 Maloney, 2008; Qi, Chantler, Siebert & Dong, 2015; Hansmann-Roth & 108 Mamassian, 2017), color appearance (Gerardin et al., 2014; Rogers, Knoblauch 109 110 & Franklin, 2016) and time perception (Lisi & Gorea, 2016). The signal detection decision model allows specifying the perceptual scales in terms of the 111 signal detection parameter d' (Gerardin et al., 2014; Knoblauch & Maloney, 112 2012). 113

The aim of the present study is to estimate perceptual scales for two dimensions, the luminance elevation of the inner contour and the luminance elevation of the background. While the luminance contrast between the inner and outer contours has been tested intensively in the WCE, experiments evaluating the influence of the background luminance are scarce. Indeed, the

WCE has generally been demonstrated for a background of higher luminance 119 than both inner and outer contours. Although the surround (e.g., the 120 background) is known to be an important influence of color appearance 121 (Brenner & Cornelissen, 2002; Brown & MacLeod, 1997; Shevell, 1978; 122 Walraven, 1976), it has not been systematically explored for the coloration 123 effect in the WCE. In addition, most studies of the WCE focus solely on its 124 coloration effect. Here, we also investigate the influences of the background 125 126 and inner contour luminances on the perceived brightness of the interior region. In summary, we employed conjoint measurement to study how both the 127 background and the inner contour luminances influence judgments of both the 128 hue and brightness in the WCE. 129

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#### 133 2. General Methods

134 2.1. Observers

Four observers participated in these experiments. Three were naïve and the fourth was one of the authors. Observers ranged in age between 26 and 40 years. All had normal color vision as tested with the Farnsworth Panel D15, and had normal or corrected-to-normal visual acuity. Experiments were performed in accordance with the principles of the Declaration of Helsinki for the protection of human subjects.

#### 142 *2.2. Apparatus*

Stimuli were presented on a NEC MultiSync FP2141sb color CRT monitor 143 driven by a Cambridge Research ViSaGe graphic board with a color resolution 144 of 14 bits per gun (Cambridge Research Systems, Rochester, United Kingdom). 145 The experimental software was written to generate all stimuli, control stimulus 146 collect responses 147 presentation and in MATLAB 7.9 (MathWorks, http://mathworks.com), using the CRS Toolbox extensions. The monitor was 148 calibrated using an OptiCal photometer with the calibration routines of 149 Cambridge Resarch Systems. Observer position was stabilized by a chinrest and 150 observer-to-screen distance was 80 cm. Experiments were performed in a dark 151 room. Both eyes were used for viewing. 152

153

## 154 *2.3. Stimuli*

The stimuli were constructed as Fourier descriptors (Zahn & Roskies, 1972). Each stimulus was defined with respect to a circle of 3.2 deg diameter whose radius, *r*, was modulated sinusoidally as a function of angle according to the equation:

159

$$R(\theta) = r + A\sin(2\pi f\theta) \tag{1}$$

where *R* is the stimulus radius at angle  $\theta$ , *r* the average radius of the stimulus, *A* the modulation and *f* the frequency in cycles per revolution (cpr). In the present study, the frequency was fixed at *f* = 10 cpr and the amplitude of both contours at *A* = 0.36 (Figure 1b, left). All stimuli were composed of three colors: an orange inner contour (x,y = 0.44, 0.43) with the luminance varying from 30.02 cd/m<sup>2</sup> to 62.74 cd/m<sup>2</sup> and a purple outer contour (x,y = 0.31, 0.11; Y = 21.12 cd/m<sup>2</sup>), presented on a neutral white background (x,y = 0.29, 0.32) with the luminance of the background (both outside and inside of the contours) varying between 35.56 cd/m<sup>2</sup> and 65.56 cd/m<sup>2</sup>. The contour pairs were each of width 16 arcmin, i.e., 8 arcmin for the inner and outer contours, each.

Stimuli were specified in the DKL color space (MacLeod & Boynton, 171 1979; Krauskopf, Williams & Heeley, 1982; Derrington, Krauskopf & Lennie, 172 1984). DKL color space is a three-dimensional opponent-modulation space 173 based on the Smith and Pokorny (1975) cone fundamentals. The sum of L and 174 M cone excitations varies on one axis (luminance), while M cone excitation 175 subtracted from L cone excitation varies on the second axis (L - M); and the sum 176 of L and M cone excitations subtracted from S cone excitation varies on the 177 third axis (S - (L + M)). The DKL axes were scaled between -1 and 1, where +/-178 1 corresponds to the maximum contrast for each axis on the monitor. The 179 stimuli were specified with the purple and orange contours at azimuth of 320 180 and 45 deg respectively. Luminance of the independent variables is specified as 181 elevation from the equiluminant plane. The luminance elevations of the orange 182 contour in DKL color space varied from -0.6 to 0 while the luminance elevation 183 of the background ranged from -0.5 to 0. 184

In the present study, five levels of luminance inner contour and five levels of luminance background were used. All levels were crossed creating a  $5 \times 5$ grid with a total of 25 stimuli. Figure 2 shows an example of the range of stimuli used, with the inner contour luminance varying across rows and the background luminance across columns.

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- 191

Figure 2 about here

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193 Control stimuli were also tested consisting of patterns that were identical 194 to the test stimuli except that the contours were interlaced (Figure 1b, right) and 195 generated little filling-in, as previously demonstrated (Devinck & Knoblauch, 196 2012; Devinck et al., 2014a; Gerardin et al., 2014). These control stimuli were 197 used to verify that observers responded to the filling-in appearance and not to 198 other stimulus features.

199

200 *2.4. Procedure* 

On each trial, two different stimuli chosen randomly from the  $5 \times 5$  grid were presented in succession to the observer. Observers performed two tasks in separate randomly ordered and counter-balanced sessions in which they compared the interior regions of the two successively presented stimuli. In the first task, observers were instructed to judge which central region evoked the strongest orange hue. In the second task, observers were asked to judge which central region appeared brighter. An equal number of test and control stimuli were interleaved in each session. With 5 levels along each of the dimensions varied, there are (25 \* 24)/2 = 300 unordered pairs. Stimuli were randomly ordered for each presentation. On each trial, a randomly chosen pair of test or control stimuli was presented. A session consisted of the random presentation of all 600 test and control pairs. Each task was repeated five times, yielding 1500 test and 1500 control trials for each observer.

214 Prior to the experiment, observers were dark-adapted for 3 min. At the beginning of each trial, a fixation cross was presented in the center of the screen 215 of duration 500 ms. At its extinction, the first pattern was presented during 500 216 ms followed by a fixation cross for 500 ms, and then the second pattern for 500 217 ms., followed by a blank screen. The observer's response initiated a 1 s pause 218 before the start of the next trial. An initial practice block of 10 trials preceded 219 220 the experiment. The experimental session started, when the observer felt at ease with the task, otherwise additional practice sessions were run. A free viewing 221 procedure was used to ensure that observer's judgments were based on foveal 222 views of the stimuli. 223

224

225 *2.5. Model* 

The data were analyzed as a decision process within the framework of a signal detection model and fit by maximum likelihood (Ho et al., 2008; Knoblauch & Maloney, 2012). Three nested models of the decision process are fit to obtain

the best prediction of the set of observers' choices: an independence model, an 229 additive model and a saturated model. Each model yields estimates of 230 perceptual scale values or internal responses that have the property that equal 231 differences in response are perceptually equal. The independence model fits the 232 observer's judgments based on only one of the component dimensions. The 233 additive model fits the judgments based on the sum of component psychological 234 responses generated by the physical dimensions. The saturated model fits the 235 236 observer's judgments including an interaction term that depends on the specific levels of the two components in addition to their simple additive combination. 237 The three models are then evaluated using a nested likelihood ratio test. This is 238 done separately for the experiments based on hue, and brightness judgments. 239 The formal description of the model is described next and follows similar 240 descriptions elsewhere (Gerardin et al., 2014; Ho et al., 2008; Knoblauch & 241 242 Maloney, 2012).

We represent the stimulus levels along the two dimensions by a variable  $\phi_{i,j}$ , where *i* and *j* correspond to the luminance levels of the background and the inner contour, respectively. In the decision models, each of the dimensions contributes a response,  $\psi_i^1$ ,  $\psi_j^2$ , to the intensity of the perceived filling-in depending on the corresponding physical intensity levels, where the superscripts correspond to the responses to the background and interior contour luminances, respectively. In the additive model, when observers judge which central area is

the more saturated orange color or which appears to be brighter, we suppose thatthe filling-in response is the sum of the component responses

252 
$$\psi_{i,j} = \psi_i^1 + \psi_j^2$$
. (2)

253 Observers compare the two presented central surfaces and the difference 254 between the filling-in strength of the stimulus  $\phi_{i,j}$  and the stimulus  $\phi_{k,l}$  is 255 computed as follows

256 
$$\Delta(i, j, k, l) = (\psi_i^1 + \psi_j^2) - (\psi_k^1 + \psi_l^2) + \varepsilon$$
(3)

where  $\varepsilon$  refers to additive noise in the decision process and is modeled as a 257 Gaussian random variable with  $\mu = 0$  and variance  $= \sigma^2$ . In plots, we indicate 258 the stimulus level by the index and not by the physical units, allowing both 259 dimensions to be plotted together. With 5 levels along each dimension, there are 260 2 \* 5 levels plus 1 variance = 11 parameters to estimate. To make the model 261 identifiable, however, the response at the lowest level along each dimension is 262 arbitrarily set to 0,  $\psi_1^1 = \psi_1^2 = 0$ , and the variance is fixed to 1 for each 263 estimated value, yielding only 8 parameters to estimate. The parameter values, 264  $\psi_i^j$  are chosen to maximize the likelihood,  $\mathcal{L}(\Psi; R)$ , over the ensemble of 265 choices, *R*, made by the observer. 266

267 
$$\mathcal{L}(\Psi; R) = \prod_{i} \Phi\left(\frac{\Delta_{i}}{2}\right)^{R_{i}} \left(1 - \Phi\left(\frac{\Delta_{i}}{2}\right)\right)^{1 - R_{i}}, \quad (4)$$

where  $\Phi$  is a cumulative distribution function for a Gaussian with mean 0 and variance 1. The 2 in the denominator of the argument scales the variance for each value of  $\psi$  to 1 so that the perceptual scale values are parameterized in terms of *d*'. In practice, this is performed using a Generalized Linear Model
with the MLCM package in R (Knoblauch & Maloney, 2012, 2014).

If the observer's judgments depend on only one of the component dimensions, we obtain the independence model, reducing the decision variable to

276 
$$\Delta(i,j,k,l) = \psi_i^1 - \psi_k^1 + \varepsilon \quad (5)$$

where the judgments depend on only dimension 1, here. In this reduced model, the values of  $\psi_j^2$  are fixed at 0 and there are only 4 free parameters to estimate. Replacing the superscript 1 by 2 yields the independence model for the other dimension.

# Finally, the saturated observer model includes an interaction factor that depend on the intensity levels of both dimensions; the decision variable is defined as follows:

284 
$$\Delta(i,j,k,l) = (\psi_i^1 + \psi_j^2 + \psi_{ij}^{12}) - (\psi_k^1 + \psi_l^2 + \psi_{kl}^{12}) +$$

Due to the interaction terms, the responses cannot be explained by a simple additive combination of components as in the previous two models. With 5 levels along each dimension, only one cell in the 5x5 grid of responses is fixed at 0 leading to 24 (the maximum) free parameters to estimate, which is the origin of the term saturated.

We analysed the data using the MLCM package (Knoblauch & Maloney,
2012, 2014) in the open source software R (R Core Team, 2017) to estimate the

(6)

ε

perceptual scale values and model the contribution of both dimensions. The likelihood ratio tests were evaluated using a  $\chi^2$  statistic with degrees of freedom the difference in number of parameters fit for each pair of models.

295

#### 296 **3. Results**

Judgments based on color and on brightness from the observers are shown in 297 Figure 3 for test and control conditions in Conjoint Proportion Plots or CPP (Ho 298 et al., 2008; Knoblauch & Maloney, 2012). In the CPP, the raw data are 299 presented in a grid format in which each cell of the grid corresponds to one 300 stimulus pair comparison. Each CPP contains all stimulus combinations and 301 summarizes the proportion of times the stimulus  $S_{kl}$  was judged for one response 302 criterion, hue (a) or brightness (b), to show a greater filling-in than the stimulus 303  $S_{ii}$ , coded according to the grey levels indicated by the color bar presented on the 304 right side of each graph. The levels of both dimensions are represented along 305 306 each axis where the 5×5 outer check indicates the stimulus levels along one dimension and with each outer check subdivided into smaller 5×5 checks 307 indicating the stimulus levels for the second dimension. Figure 3c shows the 308 expected pattern of responses for an ideal observer who chooses only the higher 309 level along one of the two stimulus dimension. The CPP presented on the left 310 311 side indicates the results when the judgments depend on the first dimension alone (here, the background luminance) and the CPP displayed on the right side 312

when the judgment depend on the second dimension (here, the inner contourluminance).

Results from the hue and brightness judgments for each observer are 315 shown in Figure 3a and 3b respectively with the results displayed on the top row 316 for the test condition and on the bottom row for the control condition in each 317 figure. For the hue judgments, the CPP for the test stimuli resembles more 318 closely the ideal CPP displayed for the second dimension, suggesting that the 319 luminance of the inner contour contributed more strongly to the choices than the 320 luminance of the background. For the brightness judgments, the CPP for the test 321 stimuli is more similar to the ideal CPP displayed for the first dimension, 322 indicating that the luminance of the background contributed more strongly to the 323 choices in comparison with the luminance of the inner contour. Deviations from 324 the ideal patterns, however, indicate contributions from both dimensions for 325 both tasks. 326

In these experiments, observers were instructed to judge the appearance of the interior region of the stimulus. However, it is possible that observers attended to the experimental dimensions (e.g., the continuity of the color of the contour) instead of the appearance of the interior region. If this were the case, we should obtain the same pattern of responses between the test and the control conditions. However, the test CPP patterns are different, with the control CPP patterns showing little, if any, systematic structure.

335

## Figure 3 about here

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Statistical comparisons of additive and independence models based on a 337 nested likelihood ratio test are shown in Table I for judgments based on hue and 338 in Table II for judgments based on brightness. In these tables, each row 339 indicates the test for an observer, and the last column corresponds to the 340 probability that the additive model fits no better than the independence model. 341 The degrees of freedom are obtained from the difference of the number of 342 coefficients estimates in each model (8 (additive) - 4 (independence) = 4). 343 Comparing the independence with the additive model indicates that the 344 independence model can be rejected for the test stimuli for both tasks and for all 345 observers (p < 0.001). The motivation for testing the nested decision models for 346 the control stimuli is less clear. Instead, we used a linear mixed-effects model to 347 test if the estimated perceptual scale values depended significantly on the 348 stimulus level with a random observer intercept (Bates et al., 2015). For both 349 the hue and the brightness judgments, no significant dependence was found 350 (hue:  $\chi^2(30) = 38.5, p = 0.14$ ; brightness:  $\chi^2(30) = 22.2, p = 0.85$ ). It could 351 be argued that 4 observers is not sufficient to estimate the variances accurately 352 in a mixed-effect model. We also performed the tests using a linear model with 353 Observer entering as a fixed-effect, with no significance obtained in either case. 354 The advantage of the mixed-effects model is that the results generalize to the 355

356	population rather than just the sample of 4 observers tested (Knoblauch &	¢
357	Maloney, 2012; Moscatelli et al., 2013; Pinheiro & Bates, 2000).	

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- 359

## Tables I and II about here

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Figure 4 shows the estimated contributions of each dimension obtained 361 from fitting the additive model for each of the tasks. For judgments based on 362 hue, the average estimated scales for each pair of inner contour luminance and 363 background luminance elevation are shown in Figure 4 (a) for 4 observers. In 364 this figure, the column labels indicate the observer identification. The top row 365 shows the scale values estimated for the test stimulus and the bottom for the 366 control stimulus. Black circles indicate the inner contour contribution and white 367 the background contribution to the judgments. The values on the abscissa 368 indicate the five stimulus levels for each dimension coded from 1 to 5. These 369 values are indices to the 5 luminance elevations of the inner contour and the 370 background used in the experiment. 371

The results in Figure 4a indicate that the contribution of the luminance of the interior contour dimension to the hue filling-in strength of the inner contour as does the contribution of the background increases with luminance elevation but with a smaller effect. The second row shows results obtained for the control stimuli. Here, there appears no systematic influence of either dimension on the strength of the coloration effect, because both dimension contributions are close

378	to zero at all stimulus levels. This result further supports that the observers
379	based their judgments on the perceived filled in color of the interior rather than
380	the luminance of the inner contour or the background.

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- 382

# Figure 4 about here

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Figure 4 (b) shows the average estimated scales for each observer for each 384 pairing of inner contour and background luminance elevations when judgments 385 were based on the brightness of the interior region. The information in the 386 figure is organized in the same fashion as Figure 4a. The top row shows the 387 additive model fits to judgments of the brightness of the interior region. As the 388 inner contour increases in luminance, the estimated contribution for this 389 component decreases, indicating that the brightness of the interior region 390 decreases, i.e., generating a relative contrast rather than an assimilative effect. 391 The background dimension contributes positively with the luminance elevation 392 to judgments of brightness. These results contrast with the scales estimated for 393 the control stimuli that show no difference and little effect of both dimensions 394 on the strength of brightness filling-in. 395

Temporal forced choice experiments can be subject to order effects (Yeshurun et al., 2008). To test for this, we compared the scales obtained for trials in which the higher luminance was presented in the first interval with the trials in which it was presented in the second interval. We performed the same

analysis with respect to the luminance of the inner contour. For the hue 400 judgments, the average estimated scales for each background luminance order 401 are shown in Figure 5 (a) for 4 observers, and judgments based on brightness are 402 displayed in Figure 5 (b). In these figures, the column labels indicate the 403 observer identification. The top row shows the estimated scale values for the 404 contribution of the inner contour luminance and the bottom for the background 405 dimension. White circles indicate the dimension contribution to the judgments 406 407 when the luminance elevation of the background is higher in the first stimulus presentation and black circles the contribution to the judgments when the 408 luminance elevation of the background is higher in the second presentation. The 409 values on the abscissa indicate the five levels of each dimension coded from 1 to 410 5. For judgments based on hue, the results in Figure 5a indicate that the 411 contribution of the inner contour dimension is not different between both 412 background presentation orders. Moreover, the background dimension is 413 approximately the same between the presentation orders. Similar results were 414 obtained when judgments were based on brightness (Figure 5b). This shows 415 that any order effect based on the luminance background is very slight or absent 416 417 in our experiments.

For completeness, we show in Figure 6 the effect of the presentation order of the luminance of the inner contour on the estimated scales for both judgments organized in the same fashion as Figure 5. For both tasks, the scales for neither dimension were influenced as a function of the presentation order.

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423

## Figure 5 and 6 about here

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Comparing the saturated and additive models with nested likelihood ratio 425 tests rejected the hypothesis that the saturated model provided a fit no better 426 then the additive model, thus, demonstrating that an interaction term is required 427 to describe the observers' judgments for both tasks (Tables III and IV). The 428 429 degrees of freedom indicate the difference of the number of coefficents estimated in the 2 models (24 (saturated) - 8 (additive) = 16). The additive 430 model fit was rejected in all 8 tests. The estimated coefficients for the saturated 431 model are shown for each observer and both tasks in the panels of Figure 7. In 432 these displays, the estimated scale values are plotted as a function of the 433 stimulus index for the luminance of the inner contour with the index of the 434 background luminance specified as a parameter for each curve. The results for 435 the hue judgments are presented on the top row with the brightness judgments 436 on the bottom. The averages of the four observers for both judgment conditions 437 are summarized as three-dimensional surfaces in Figures 8a and b. If the 438 additive model was a good fit to the data, the curves for different background 439 levels would be parallel in Figure 7. However, for both the hue and brightness 440 responses, as the background luminance increases, the curves fan out. For the 441 hue judgments, this results in a larger range of hue responses to the range of 442 inner contour luminances tested for the higher than lower luminance 443

444	backgrounds. For the brightness judgments, in contrast, the range of response
445	decreases at the highest luminance background. Thus, the background
446	luminance produces different effects on both the type of filling-in (assimilation
447	vs contrast) for hue and brightness judgments and on the dynamic range of the
448	response, expanding it for hue judgments but compressing it for brightness.
449	
450	Figures 7 and 8 about here
451	
452	Tables III and Table IV about here
453	
454	4. Discussion
455	In the present study, MLCM was used to quantify the contributions to the
456	filling-in strength of the WCE of two stimulus dimensions: the background and
457	inner contour luminances. We quantified how changes in these features affected
458	perceived filling-in using two separate response criteria, linked to the strength of
459	the hue (the conventional WCE) and the brightness of the interior region.
460	Control experiments using a stimulus contour that generated statistically
461	undetectable filling-in confirmed that observers judged the perceived attributes
462	of the interior regions and not changes in the background and contour
463	luminances, per se. We also found that the results were largely independent of
464	the ordering of presentation in a trial of both the backgrounds and the inner

As found previously, the strength of the coloration effect depends on the 466 luminance of the inner contour (Devinck & Knoblauch, 2012; Devinck et al., 467 2014; Gerardin et al., 2014). Here, we show that the stimulus configuration 468 inducing the WCE generates both a hue and also a brightness filling-in of the 469 interior area and that these two phenomena are differentially affected by the 470 stimulus dimensions that we manipulated. The hue filling-in effect was 471 assimilative, and the hue became more saturated with increases in luminance of 472 the inner contour and the background. However, luminance of the interior 473 contour generated a contrast effect for brightness, in that the judged brightness 474 of the interior region decreased at higher contour luminances. As observers 475 compared the interior regions of the two successively displayed stimuli and 476 judged which central region appeared brighter, it is still possible that the 477 perceived filling-in was assimilative, i.e., the same contrast polarity as the inner 478 contour and that the effect of the contour luminance was simply to reduce the 479 lightness. This is difficult to assess since we cannot simply compare the interior 480 with the surround because both the surround and the interior vary in each 481 condition. Additionally, if we hold the surround constant, then we cannot rule 482 out its effect on the interior. 483

484 The hue effect

Both the hue and the brightness of the interior region were judged to be greater with increases in the luminance elevation of the background. Pinna et al. (2001) have previously reported that color spreading in the WCE occurs not

only with a white background but also with grey backgrounds and that even a
faint spreading is perceived with dark backgrounds. Our results agree with these
observations in that the background contribution to the hue judgments increased
at higher luminances.

We also found that hue filling-in was more pronounced for bright than 492 dim backgrounds. Previous studies have reported that color saturation 493 diminishes when the brightness contrast between a colored object and its 494 495 luminance background increases, a phenomenon named the gamut expansion effect (Brown & McLeod, 1997) subsequently confirmed and quantified by 496 several investigators (Bimler et al., 2009; Faul et al., 2008; Xing et al., 2015). 497 In contrast with these studies, our experiment shows that increasing the 498 background luminance strengthens the assimilation hue, suggesting that we are 499 observing a different phenomenon. 500

## 501 The brightness effect

The WCE has typically been described as a coloration effect. Brightness variations in the interior region of the stimulus have not typically been systematically quantified. One exception to this concerns studies that used an achromatic stimulus configuration (Cao et al., 2011; Coia & Crognale, 2017). In Coia and Crognale (2017), observers compared the filling-in region in the WCE to a reference stimulus with a physical luminance difference. Their matching results indicated that the test field shifted in the opposite direction

from the inner contour showing an assimilation effect. In Cao et al. (2011), the 509 luminances of the inner contour and the luminance of the background were fixed 510 while the outer contour varied between high and low luminance levels, thus 511 varying the contrast. Observers were asked to report which of two interior 512 surface stimuli appeared darker. Their results followed a U-shape with the 513 strength of the effect maximized for a range of medium luminance levels but not 514 for the extreme luminance levels. Consequently, the luminance contrast 515 516 between both contours affects the WCE but not linearly. An intriguing point is the authors' description of their results: "Although there is an apparent 517 assimilation effect in the chromatic WCE, it is hard to tell whether it is actually 518 assimilation or some type of contrast effect happening here for the 'opposite 519 polarity' condition". As their method based the estimation of the WCE strength 520 on the probability of discriminating which stimulus appeared darker, we can 521 assume that if responses are inferior to 50%, then the surface appears lighter but 522 not equal (which would be the case for responses equal to 50%), indicating a 523 contrast rather than an assimilation effect. For our experiment based on 524 brightness judgment, when the inner contour increases in luminance, the 525 estimated contribution for this component decreases, indicating that the 526 brightness of the interior region decreases, also indicating a contrast 527 phenomenon. Thus, the two sets of results may be in agreement. Nevertheless, 528 the nature of the brightness effect could be influenced by the presence of the 529 chromatic component in our stimulus situation. 530

In Devinck et al. (2005), observers were allowed to adjust both the 531 luminance and the chromaticity of a field in order to match the color of the 532 WCE. The mean luminance match was near the luminance of the background, 533 indicating that little or no luminance adjustment was required to make a 534 perceptual match. The authors concluded that the WCE is predominantly a 535 chromatic effect as originally suggested by Pinna et al. (2001). These results are 536 not necessarily in conflict with the current study, in that we observed that the 537 variation in brightness contrast with inner contour luminance is diminished at 538 high backgrounds and might have been difficult to detect via matching. We 539 would predict, then, that manipulating the luminance of the background would 540 affect the luminance match, but this would require a more systematic study of 541 the background than was performed by Devinck et al. (2005). Here, our data 542 indicate that the perceptual effect is not limited only to a coloration phenomenon 543 in that both the background and inner contour luminances influence observers' 544 judgment of the brightness of the central surface. A simple hypothesis to 545 account for the reduced brightness effect at high background luminances is to 546 suppose that the background light added to the interior region cancels the 547 contrast effect. This would not explain, however, the fact that higher 548 background luminances led to a stronger hue percept. 549

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#### 553 Assimilation vs contrast

Whether one observes a contrast or assimilation effect may depend on the width 554 of the contours (Fach & Sharpe 1986; Helson & Rohles, 1959; Helson, 1963). 555 Other factors such as the luminance of the inducing stimuli can also influence 556 our percept. Thus, de Weert and Spillmann (1995) indicated that assimilation or 557 contrast occurred depending on whether the inducing contours of varied 558 reflectance were darker or lighter than the gray background in using a 559 pincushion pattern. A matching experiment for brightness judgements indicated 560 that contrast occurred when the luminance level of the inducing contour was 561 above the luminance level of the background and that assimilation occurred 562 when the luminance of the inducing contour was below the luminance of the 563 Given the different spatial dependencies of chromatic and background. 564 luminance sensitive mechanisms, one might expect differences in the spatial 565 domains over which chromatic and luminance components of a stimulus induce 566 assimilation and contrast. Given the dependence of induction phenomena on 567 stimulus configuration, however, (Fach & Sharpe 1986; de Weert & Spillmann 568 1995; Smith et al. 2001; Monnier & Shevell 2003, 2004), it is difficult to predict 569 *a priori* whether the dimensions of the contours that we used should predict one 570 or the other for the hue and brightness judgments. 571

In our study, observers compared the interior regions of the two successively displayed stimuli and judged which central region appeared brighter. Under this condition, our results indicated that the brightness of the

interior region decreases generating a relative contrast rather an assimilation 575 effect. However, it still possible that different visual phenomena are perceived 576 in other circumstances. In making a brightness judgements of the central region 577 with respect to the outer region, we noted an assimilation phenomenon at the 578 lowest luminance level of the background and at the highest luminance level of 579 the inner contour. This condition corresponds to the lower left corner in Figure 580 2. 581

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#### Unitary vs multiple mechanisms

The color of the central surface in the WCE is characterized by a spread 583 of color from the inner contour. Most previous studies of the WCE reported that 584 the coloration effect depends on both the chromatic and luminance contrasts of 585 the inner and outer contours. For example, most authors demonstrated that the 586 coloration effect increases with increasing luminance contrast between inner and 587 outer contours (Devinck & Knoblauch, 2012; Devinck et al., 2005). 588 Additionally, the coloration effect increases when the chromatic coordinates of 589 the inner and outer contours are approximately complementary in the color 590 diagram (Devinck et al., 2006). Thus, the main explanation assumes a filling-in 591 process in which a neuronal mechanism detects the contour and generalizes it 592 beyond the confines of the immediate stimulus. Most studies in the WCE have 593 reported an important role for several types of contour mechanisms generating a 594 long-range filling-in percept. Taken together, these data suggest that the filling-595 in process involved in the WCE requires multiple levels of processing (Devinck 596

et al., 2014b; Pinna et al., 2001; Pinna & Grossberg, 2005; von der Heydt & 597 Pierson, 2006). The present results indicate that the mechanisms inducing the 598 brightness and coloration percept in the WCE are affected differently by the 599 luminance of the inner contour. These opposing responses due to the inner 600 contour suggest that multiple mechanisms contribute to the appearance of the 601 interior region. Different mechanisms could be activated or inhibited yielding to 602 color assimilation or brightness contrast effects, respectively. 603 Future 604 experiments based on visual masking or contour adaptation could investigate such phenomena. 605

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726 727 728 729 730 731	

# 732 Figure Legends

733	Fig. 1:	(a) Example of the Watercolor Effect. When a light orange contour is
734		surrounded by a dark purple contour, the enclosed area takes the tint of
735		the orange border. (b) Example of stimuli using Fourier descriptor as
736		test stimulus (presented on the left side) and using braided contour as
737		control stimulus (displayed on the right side).

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Fig. 2: Examples of stimulus set used for a conjoint measurement experiment.
The figure indicates the set of stimuli used in both judgment tasks.
Each column corresponds to a different luminance elevation of the
background and each row to a different luminance elevation of the
inner orange contour.

744

Conjoint proportion plots for judgments based on hue (a) and on Fig. 3: 745 brightness (b). Each plot shows the proportion of stimulus  $S_{kl}$  judged to 746 have a greater filling-in than the stimulus represented in abscissa  $S_{ii}$  as 747 grey level according to the color bar on the right side. The luminance 748 elevation of the background is indicated by the large grids (i,k) and 749 each grid is subdivided into smaller 5×5 grid indicating the luminance 750 elevation of the inner contour (j,l). In each set of graphs, the top row 751 indicates the results for the test stimuli and the bottom for the control 752

stimuli for 4 observers. (c) Conjoint proportion plots for a simulated
observer. Expected responses for an observer who judges the stimuli
based on the contributions along only one of the dimensions.

756

Fig. 4: (a) Estimated scales for judgments based on hue. Additive model 757 average estimates for test (top row) and control stimuli (bottom row) as 758 a function of inner contour elevation (black circles) and luminance 759 760 background elevation (white circles) for four observers. Error bars show 95% confidence intervals for estimates across the 5 runs. (b) 761 Results for judgments based on brightness. The solid lines indicate the 762 estimated contributions of each dimension under the additive model for 763 test (top row) and control patterns (bottom row) as a function of inner 764 contour elevation (black circles) and luminance background elevation 765 (white circles) for four observers. Error bars show 95% confidence 766 intervals for estimates across the 5 runs. 767

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Fig. 5: (a) Results for judgments based on hue depending on the presentation
order for the luminance background of the test stimulus. The solid
lines indicate the estimated contributions under the additive model for
the inner contour dimension (top row) and the background dimension
(bottom row) as a function of the background order presentation. The
white circles are used to indicate that the luminance elevation of the

background is higher in the first interval than in the second interval and 775 the black circles are used to represent a higher luminance elevation of 776 the background in the second interval than in the first interval. Error 777 bars are 95% confidence intervals based on a bootstrap procedure 778 (Knoblauch & Maloney, 2012). (b) Results for judgments based on 779 brightness depending of the luminance background presentation order 780 for the test stimulus. Additive model average estimates for the inner 781 782 contour dimension (top row) and the background dimension (bottom row) when the luminance elevation of the background is higher in the 783 first interval than in the second interval (white circles) and when the 784 luminance elevation of the background is higher in the second interval 785 than in the first interval (black circles). Error bars are 95% confidence 786 intervals based on a bootstrap procedure (Knoblauch & Maloney, 787 2012). 788

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Fig. 6: Results for judgment based on hue (a) and on brightness (b) depending
on the presentation order for the luminance contour of the test stimulus.
The information in the figure is organized in the same fashion as Figure
5. The white circles are used to indicate that the luminance elevation of
the inner contour is higher in the first interval than in the second
interval and the black circles are used to represent a higher luminance

796	elevation	of the	inner	contour	in	the	second	interval	than	in	the	first
797	interval.	Error b	ars are	e 95% co	nfic	denc	e interv	als.				

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Fig. 7: Results of the estimated contributions for each combination of the two
dimensions under the saturated model for four observers. The different
lines are used to code the index of the background dimension (for
indices 1 to 5). The top row represents the estimated contribution when
judgment is based on hue and the bottom row is the estimated
contribution when judgment is based on brightness.

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Fig. 8 Three-dimensional surfaces for the average of the data over observers from Figure 7, showing the contributions of the inner contour and background to a) the hue and b) the brightness judgments under the saturated model.

(a)



(b)



Figure 1



# Background Luminance

Figure 2







(b)



Stimulus S<sub>ii</sub> Background Level (i) and Inner Contour Level (j)

(c)









(b)













(b)

> З



Physical Scale Index

Figure 6



Physical Scale Index (Interior Contour Luminance)

Figure 7



Figure 8

Table I: Comparison of independence and additive models for judgements based on hue

	Test stime	Test stimuli (additive model × independent				
		model)				
	Df	Deviance	p-value			
Obs. 1	4	270.22	< 0.001			
Obs. 2	4	568.41	< 0.001			
Obs. 3	4	309.60	< 0.001			
Obs. 4	4	406.38	< 0.001			

Table II: Comparison of independence and additive models for judgement based on brightness

	Test stimuli (additive model × independent					
		model)				
	Df	Deviance	p-value			
Obs. 1	4	275.434	< 0.001			
Obs. 2	4	912.11	< 0.001			
Obs. 3	4	400.82	< 0.001			
Obs. 4	4	430.64	< 0.001			

Table III: Comparison of additive and saturated models for judgement based on hue

	Test stimuli (additive model × saturated model)				
	Df	Deviance	p-value		
Obs. 1	16	105.44	< 0.001		
Obs. 2	16	142.82	< 0.001		
Obs. 3	16	104.81	< 0.001		
Obs. 4	16	108.14	< 0.001		

<i>Table IV</i> : Comparison of additive and	saturated models for judgement b	based on brightness
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	Test stimuli (additive model × saturated model)				
	Df	Deviance	p-value		
Obs. 1	16	56.154	< 0.001		
Obs. 2	16	74.01	< 0.001		
Obs. 3	16	32.83	< 0.01		
Obs. 4	16	122.78	< 0.001		