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6	A smoothness constraint on the development of object recognition			
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16	Word Count: 2,974			
17				

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#### Abstract

19 Understanding how the brain learns to recognize objects is one of the ultimate goals in the 20 cognitive sciences. To date, however, we have not vet characterized the environmental factors 21 that cause object recognition to emerge in the newborn brain. Here, I present the results of a 22 high-throughput controlled-rearing experiment that examined whether the development of object 23 recognition requires experience with temporally smooth visual objects. When newborn chicks 24 (Gallus gallus) were raised with virtual objects that moved smoothly over time, the chicks 25 developed accurate color recognition, shape recognition, and color-shape binding abilities. In 26 contrast, when newborn chicks were raised with virtual objects that moved non-smoothly over 27 time, the chicks' object recognition abilities were severely impaired. These results provide evidence for a "smoothness constraint" on newborn object recognition. Experience with 28 29 temporally smooth objects facilitates the development of object recognition.

### 31 **1. Introduction**

32 Object recognition is one of the most important functions of the vertebrate visual system. 33 To date, however, the development of object recognition is poorly understood. What 34 environmental factors cause object recognition to emerge in the newborn brain? Does this ability 35 emerge automatically, or do newborns require a specific type of visual input in order to develop 36 accurate object recognition abilities? These types of questions are difficult to address with 37 humans because human infants cannot be raised in strictly controlled environments from birth. In 38 contrast, questions that concern the role of experience in development can be addressed directly 39 with controlled-rearing studies of newborn animals. Here, I describe a high-throughput 40 controlled-rearing experiment that examined whether the development of object recognition 41 requires experience with temporally smooth visual objects.

42 Researchers have long theorized that biological visual systems leverage the temporal 43 smoothness of natural visual environments to recognize objects (e.g., DiCarlo, Zoccolan, & Rust, 44 2012; Feldman & Tremoulet, 2006; Foldiak, 1991; Gibson, 1979; Stone, 1996; Wallis & Rolls, 45 1997; Wiskott & Sejnowski, 2002). In particular, when an object moves smoothly across the 46 visual field, the object projects a series of gradually changing images on the retina. The visual 47 system might take advantage of this natural tendency for temporally contiguous retinal images to 48 belong to the same object by associating patterns of neuronal activity produced by successive 49 retinal images of an object. When provided with temporally smooth visual input, this temporal 50 association process should create object representations that are selective for object identity and 51 tolerant to identity-preserving image transformations (e.g., changes in viewpoint).

52 A wealth of studies provide evidence that mature visual systems use temporal association 53 mechanisms to create object representations. For example, when human adults are presented with

54 objects that rotate in depth, sequential views come to be associated with one another in a manner 55 that aids recognition (Cox, Meier, Oertelt, & DiCarlo, 2005; Liu, 2007; Stone, 1998; Vuong & 56 Tarr, 2004; Wallis, Backus, Langer, Huebner, & Bulthoff, 2009; Wallis & Bulthoff, 2001). 57 Temporal association effects have also been found on the neurophysiological level in adult 58 monkeys (Li & DiCarlo, 2008, 2010; Meyer & Olson, 2011; Miyashita, 1988). In the present 59 study, I examined whether newborn visual systems create more accurate object representations 60 when presented with temporally smooth objects compared to temporally non-smooth objects—as 61 predicted by temporal association models (Wallis, 1998; Wallis & Bulthoff, 2001). Specifically, 62 I examined the *first* visual object representation created by newborn subjects, before their visual 63 systems had been shaped by any prior visual object experience.

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### 65 1.2. A high-throughput controlled-rearing method

This experiment required controlling all of the subjects' visual experiences from the onset of vision and measuring their object recognition abilities across a range of test trials. To meet these requirements, I used a high-throughput controlled-rearing method (Wood, 2013). The method involves raising newborn chicks in strictly controlled environments and recording their behavior in response to pre-programmed animations (Figure 1A). We use the term "highthroughput" to describe the method because the controlled-rearing chambers record all of the subjects' behavior (24/7).

I used domestic chicks as an animal model because they are an ideal model system for studying the development of vision (Wood & Wood, 2015a). First, chicks can be raised in strictly controlled environments immediately after hatching, which makes it possible to control all of their visual object experiences. Second, chicks imprint to objects seen in the first days of

life. This imprinting behavior can be used to test chicks' object recognition abilities without 77 78 training (Bateson, 2000; Horn, 2004). Third, birds and mammals process sensory input using 79 homologous neural circuits with similar connectivity patterns (reviewed by Jarvis et al., 2005; 80 Karten, 2013). Since birds and mammals use homologous neural mechanisms to process visual 81 input, controlled-rearing studies of newborn chicks can inform our understanding of the 82 development of both avian and mammalian vision. Finally, chicks develop visual recognition 83 abilities rapidly (Vallortigara, 2012). For example, newborn chicks can begin recognizing objects 84 (Wood, 2013, 2015), faces (Wood & Wood, 2015b), and actions (Goldman & Wood, 2015) at 85 the onset of vision. Newborn chicks can also build integrated object representations with bound 86 color-shape units (Wood, 2014).

In the first week of life (input phase), newborn chicks were raised in environments that contained no objects other than a single virtual object (Figure 1A). For one group of chicks, the virtual object moved smoothly over time (Temporally Smooth Condition), whereas for another group of chicks, the virtual object moved non-smoothly over time (Temporally Non-Smooth Condition). In the second week of life (test phase), I used an automated two-alternative forcedchoice procedure to test the chicks' color recognition, shape recognition, and color-shape binding abilities.

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### 95 **2. Method**

96 2.1. Subjects

97 Twenty-two domestic chicks of unknown sex were tested. No subjects were excluded 98 from the analyses. The eggs were obtained from a local distributor and incubated in darkness in 99 an OVA-Easy incubator (Brinsea Products Inc., Titusville, FL). After hatching, the chicks were 100 moved from the incubation room to the controlled-rearing chambers in complete darkness. Each 101 chick was raised singly within its own chamber. Ten chicks were raised with a temporally 102 smooth object and 12 chicks were raised with a temporally non-smooth object.<sup>1</sup> This experiment 103 was approved by The University of Southern California Institutional Animal Care and Use

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- 105
- 106 2.2. Controlled-Rearing Chambers

Committee.

107 The controlled-rearing chambers (66 cm length  $\times$  42 cm width  $\times$  69 cm height) were 108 constructed from white, high-density polyethylene and were devoid of all real-world (solid, 109 bounded) objects. To present object stimuli to the chicks, virtual objects were projected on two 110 display walls situated on opposite sides of the chamber. The display walls were 19" liquid crystal 111 display (LCD) monitors (1440  $\times$  900 pixel resolution). Food and water were provided within 112 transparent troughs in the ground (66 cm length  $\times$  2.5 cm width  $\times$  2.7 cm height). Grain was used 113 as food because it does not behave like an object (i.e., grain does not maintain a rigid, bounded 114 shape). The floors were wire mesh and supported 2.7 cm off the ground by thin, transparent 115 beams. The chambers tracked all of the chicks' behavior (9 samples/second, 24 hours/day, 7 days/week) via micro-cameras in the ceilings and automated image-based tracking software 116 117 (EthoVision XT, Noldus Information Technology, Leesburg, VA). This high-throughput data 118 collection approach allowed us to collect a large number of test trials (168 trials) from each 119 chick, and consequently, measure each subject's object recognition abilities with high precision.

<sup>&</sup>lt;sup>1</sup> The results from the Temporally Smooth Condition were described previously in Wood (2014). In the present study, I directly contrasted chicks raised with temporally smooth objects and temporally non-smooth objects. While the chicks in the two conditions were not tested concurrently, they were tested with the same automated method. Indeed, one major benefit of this controlled-rearing method is that different groups of subjects can be tested in exactly the same way, since the stimuli presentation and data collection processes are fully automated.

In total, 7,392 hours of video footage (14 days × 24 hours/day × 22 subjects) were collected for
this experiment.

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#### 123 *2.3. Procedure*

In the first week of life (input phase), the chicks were raised in controlled-rearing chambers that contained no objects other than a single virtual object (Figure 1A). The object appeared on one display wall at a time, switching walls every two hours (Figure 2A). On average, the object measured 9 cm (length)  $\times$  7 cm (height) and was displayed on a uniform white background. Half of chicks were imprinted to the object shown in Figure 1B and half of the chicks were imprinted to the object shown in Figure 1C.

130 In the Temporally Smooth Condition (SI Movie 1), the virtual object rotated smoothly 131 around a frontoparallel vertical axis, completing a full rotation every 6 seconds (30 132 frames/second). The object had two faces, each with a different color and shape (Figure 1B). Since the edges of the object (shown during transitions from one face to the other) were identical 133 134 in color and shape, the object appeared to change smoothly from one 3-D shape to the other 3-D 135 shape. Using this type of geometrically impossible object allowed two different color-shape units 136 to be presented on a single smoothly moving object. Accordingly, I was able to examine whether 137 the first object representation built by newborn chicks contains integrated color-shape units. The 138 same temporally smooth movie was presented throughout the input phase; thus, the transitional probability between images was 1.0.<sup>2</sup> 139

 $<sup>^2</sup>$  The term "transitional probability" refers to the consistency with which the visual images occurred in a particular order. Since the images were presented in a constant order throughput the input phase, the transitional probability between images was 1.0 in both conditions.

140 In the Temporally Non-Smooth Condition (SI Movie 2), the chicks were shown the same 141 virtual object, but the object images were presented in a scrambled order (Figure 1B). 142 Specifically, I took the 180 unique images (30 frames/second  $\times$  6 seconds) from the temporally 143 smooth animations and randomized the order of the images. On average, the successive images 144 differed by 154° and the minimum difference between two successive views was 50°. To make 145 the images more distinct and eliminate flicker, each image was presented for one second. The 146 same non-smooth movie was presented throughout the input phase; thus, the transitional 147 probability between images was 1.0.

Critically, the virtual objects presented in the two conditions were composed of the same individual images and were equally predictive in terms of the transitional probabilities between images. Furthermore, the subjects received the same amount of overall time with each individual image across the conditions (despite the images being presented at different rates). Thus, any difference in recognition performance between the conditions could not be based on the amount of exposure to the individual images or the transitional probabilities between images.

154 In the second week of life (test phase), I examined whether the chicks could recognize 155 their imprinted object across a variety of feature changes. During the test trials, two objects were 156 shown simultaneously, one on each display wall (Figure 2B). One object was the imprinted 157 object from the input phase, and the other object was an unfamiliar object. If the chicks could 158 distinguish their imprinted object from the unfamiliar object, then they should have spent a 159 greater proportion of time in proximity to the imprinted object compared to the unfamiliar 160 object. The chick was considered to be in proximity to an object when the chick occupied a  $22 \times$ 161 42 cm zone next to the display wall showing that object. The test objects moved smoothly in the

Temporally Smooth Condition and non-smoothly in the Temporally Non-Smooth Condition. During the test trials, subjects were presented with the following test trial types:

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165 Color Change Trials: The imprinted object was paired with an unfamiliar object that was
 166 identical to the imprinted object except that one or both colors were replaced with novel
 167 colors.

Shape Change Trials: The imprinted object was paired with an unfamiliar object that was
 identical to the imprinted object except that one or both shapes were replaced with novel
 shapes.

- 171 *Color-Shape Change Trials*: The imprinted object was paired with an unfamiliar object in which
   172 one face was replaced with a novel color and shape or both faces were replaced with
   173 novel colors and shapes.
- Binding Change Trials: The imprinted object was paired with an unfamiliar object that had the same color and shape features as the imprinted object, but in a different configuration (e.g., a yellow triangle and a purple circle vs. a yellow circle and a purple triangle).
- 177

During the test phase, subjects received 168 test trials (1 trial per hour). Each test trial lasted 20 minutes and was followed by a 40-minute rest period. During the rest periods, the animation from the input phase was shown on one display wall, and the other display wall contained a white screen.

- 183 **3. Results**
- 184 *3.1. Recognition Performance*

185 The results are shown in Figure 3. For each test trial type, I computed the percent of time 186 each chick spent with the imprinted object compared to the unfamiliar object. A repeated 187 measures ANOVA with Test Trial Type as a within-subjects factor and Condition (Temporally 188 Smooth vs. Temporally Non-Smooth) as a between-subjects factor revealed a significant main 189 effect of Test Trial Type (F(6, 120) = 17.08, p < .001). The interaction was not significant (F(6, 120) = 17.08, p < .001). 190 120 = 0.65, p = .69). Recognition performance was significantly higher in the Temporally 191 Smooth Condition than the Temporally Non-Smooth Condition, both for overall recognition 192 performance (t(20) = 3.44, p = .003). Cohen's d = 1.54 and for each of the seven test trial types 193 (see Table 1). In brief, newborn chicks showed superior recognition performance when raised 194 with temporally smooth objects.

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## 196 3.2 Measuring the Strength of the Imprinting Response

197 One potential explanation for this effect is that the chicks imprinted more strongly to the 198 temporally smooth objects than the temporally non-smooth objects, and were therefore more 199 motivated to approach the temporally smooth objects. To examine whether temporal smoothness 200 influenced the strength of the imprinting response, I examined the proportion of time chicks 201 spent by the imprinted objects during the rest periods. During the rest periods, the imprinted 202 object was projected on one display wall while the other display wall was blank. Thus, the rest 203 periods provided a measure of the amount of time subjects generally preferred to spend in 204 proximity to their imprinted object. The chicks in the Temporally Smooth and Temporally Non-205 Smooth Conditions spent 86.0% (SEM = 1%) and 86.2% (SEM = 1%) of their time with their 206 imprinted object, respectively. These values did not differ significantly from one another (t(20) =

207 .17, p = .87). The chicks imprinted equally strongly to the temporally smooth and temporally 208 non-smooth objects.

209

### 210 **4. Discussion**

211 I used a high-throughput controlled-rearing method to examine whether newborn chicks 212 need experience with temporally smooth visual objects to develop object recognition abilities. 213 The chicks raised with the temporally smooth objects and the chicks raised with the temporally 214 non-smooth objects were exposed to the same individual images, and the objects were equally 215 predictive in terms of the transitional probabilities between images; nevertheless, there were 216 significant differences in recognition performance across the groups. When newborn chicks were 217 raised with a temporally smooth object, they developed accurate color recognition, shape 218 recognition, and color-shape binding abilities. In contrast, when newborn chicks were raised with 219 a temporally non-smooth object, their object recognition abilities were impaired. Thus, there is a 220 "smoothness constraint" on newborn object recognition. Experience with temporally smooth 221 objects facilitates the development of object recognition.

222 These results accord with previous studies showing that temporal learning abilities can 223 emerge within the first few months of life (e.g., Bulf, Johnson, & Valenza, 2011; Kirkham, 224 Slemmer, & Johnson, 2002; Kirkham, Slemmer, Richardson, & Johnson, 2007), and extend this 225 literature by showing that newborn visual systems use temporal learning mechanisms at the onset 226 of vision when building their first visual object representation. These results also accord with 227 temporal association models in two respects. First, temporal association models predict that 228 smooth changes in an object's appearance over time will result in larger changes in neural 229 selectivity (and hence, discrimination performance) than non-smooth changes in appearance

(Wallis, 1998; Wallis & Bulthoff, 2001). Consistent with these models, newborn chicks 230 231 developed enhanced object recognition abilities when raised with temporally smooth objects. 232 Second, temporal association models predict that it should be possible to create 'unnatural' 233 object representations by exposing subjects to visual worlds with unnatural spatiotemporal 234 statistics (Cox et al., 2005; Li & DiCarlo, 2008, 2010). In the present study, newborn chicks were 235 exposed to an unnatural visual world with a geometrically impossible object; nevertheless, the 236 chicks were able to build a robust representation of the object (provided that the object was 237 temporally smooth).

238 It is important to emphasize that temporal smoothness is a continuous variable rather than 239 a binary variable. Since chickens have a relatively high flicker fusion rate ( $\sim 100$  Hz), it is 240 possible that even the temporally smooth movies were not perceived as completely smooth by 241 the chicks. Temporal smoothness is also a broad term that can refer to many different types of 242 change across images. Visual sequences can be temporally smooth from a brightness perspective, 243 pixel-level perspective, feature-level perspective, and so forth. It would be interesting for future 244 studies to systematically manipulate the amount and type of temporal smoothness in the visual 245 environment, and examine the effects of those manipulations on chicks' object recognition 246 abilities.

To what extent do these findings apply to the development of object recognition in humans? In some respects, we should expect differences in the development of object recognition between chickens and humans. For instance, newborn humans are relatively immature at birth and have difficulty detecting the direction and speed of motion (Wattam-Bell, 1991, 1992). Accordingly, newborn humans might process temporally smooth visual input differently than newborn chicks. On the other hand, there is growing evidence in the 253 neurosciences for an evolutionarily ancient cortical circuit for processing sensory information 254 (reviewed by Karten, 2013). This circuit is thought to have evolved in stem amniotes at least 100 255 million years ago (Jarvis et al., 2005) and to underlie the computations used for visual object 256 recognition (DiCarlo et al., 2012). If mammals and birds share homologous neural circuits for 257 processing visual input, as these findings suggest, then human visual systems should be subject 258 to similar constraints as chicken visual systems. It would be interesting for future studies to 259 examine directly whether object recognition in human infants is subject to a smoothness 260 constraint.

In conclusion, the present study provides evidence for a smoothness constraint on the development of object recognition in a newborn biological visual system. Newborn chicks can develop color recognition, shape recognition, and color-shape binding abilities rapidly (within the first week of life), but these abilities do not emerge automatically. Rather, robust object recognition abilities emerge when newborn chicks are raised with temporally smooth objects.

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- 343
- 344

345	45 Acknowledgements			
346	This research was funded by National Science Foundation CAREER Grant BCS-1351892. I			
347	thank Brian W. Wood for assistance with the supplementary movies and Samantha M. W. Wood			
348	for helpful comments on the manuscript.			
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### **Figure Captions**

351 Figure 1: (A) Illustration of a controlled-rearing chamber. The chambers were devoid of real-352 world objects. To present object stimuli to the chicks, virtual objects were projected on 353 two display walls situated on opposite sides of the chamber. During the input phase (1<sup>st</sup> 354 week of life), newborn chicks were exposed to a single virtual object that either moved 355 smoothly or non-smoothly over time. Half of the chicks were raised with the object 356 shown in panel (**B**) and half of the chicks were raised with the object shown in panel (**C**). 357 Figure 2: The experimental procedure. These schematics illustrate how the virtual stimuli were 358 presented for sample 4-hour periods during the (A) input phase and (B) test phase. 359 During the input phase, newborn chicks were exposed to a single virtual object with two 360 faces, each with a different color and shape. The object appeared on one wall at a time 361 (indicated by blue segments on the timeline), switching walls every 2 hours, after a 1-min 362 period of darkness (black segments). During the test phase, two virtual objects (one 363 imprinted, one unfamiliar) were shown simultaneously, one on each display wall, for 20 364 minutes per hour (gray segments). The illustrations below the timeline are examples of 365 paired test objects displayed in four of the test trials. Each test trial was followed by a 40-366 min rest period (blue segments). During the rest periods, the animation from the input 367 phase was shown on one display wall, and the other display wall was blank. These 368 illustrations show the displays seen by the subjects that were imprinted to Imprinted 369 Object 1 (Figure 1). (C) The unfamiliar objects presented in the four test trials depicted in 370 Figure 2B. These illustrations show how the object transitioned from face to face in the 371 Temporally Smooth Condition.

Figure 3: Results from the test phase. The bar graph shows the mean proportion of time spent by
the imprinted object compared to the unfamiliar object for each test trial type. Error bars
denote ±1 SE. Chance performance (dashed lines) was 50%.

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Trial Type	<i>t</i> -score	<i>p</i> -value	Cohen's d
1 Color Change	2.25	.04	1.01
2 Color Changes	2.05	.05	.92
1 Shape Change	2.84	.01	1.27
2 Shape Changes	2.99	.007	1.34
1 Color-Shape Change	2.19	.04	.98
2 Color-Shape Changes	2.13	.05	.95
Binding Change	3.30	.004	1.48

# Tables

**Table 1:** Results of two-tailed, independent samples *t*-tests comparing object recognition
performance across the Temporally Smooth and Temporally Non-Smooth Conditions for the
seven trial types.





# Figure 2





Figure 3