

## 3D Printed Capacitive Sensor Objects for Object Recognition Assays

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35

36

37 **Abstract**

38 Object recognition tasks are widely used assays for studying learning and memory in rodents. Object  
39 recognition typically involves familiarizing mice with a set of objects and then presenting a novel object  
40 or displacing an object to a novel location or context. Learning and memory are inferred by a relative  
41 increase in time investigating the novel/displaced object. These tasks are in widespread use, but there  
42 are many inconsistencies in the way they are conducted across labs. Two major contributors to this are  
43 the lack of consistency in the method of measuring object investigation and the lack of standardization  
44 of the objects that are used. Current video-based automated algorithms can often be unreliable  
45 whereas manual scoring of object investigation is time-consuming, tedious, and more subjective. To  
46 resolve these issues, we sought to design and implement 3D printed objects that can be standardized  
47 across labs and utilize capacitive sensing to measure object investigation. Utilizing a 3D printer,  
48 conductive filament, and low-cost off-the-shelf components, we demonstrate that employing 3D printed  
49 capacitive touch objects is a reliable and precise way to perform object recognition tasks. Ultimately,  
50 this approach will lead to increased standardization and consistency across labs, which will greatly  
51 improve basic and translational research into learning and memory mechanisms.

52

53 **Significance Statement**

54 Object recognition assays are widely used in basic research and preclinical models; however, there is a  
55 profound lack of standardization in the objects used and scoring methods employed. Here, we show a  
56 proof-of-principle demonstration that employing 3D printed capacitive objects is a cost-effective,  
57 reliable, and precise way to perform object recognition tasks when compared to manual scoring. This  
58 novel approach could ultimately contribute to a more standardized approach to object recognition  
59 tasks, which would greatly improve reliability in basic and applied neurobehavioral research.

60

61 **Introduction**

62 Object recognition tasks are widely used assays for studying learning and memory in rodents (Antunes &  
63 Biala, 2012; Ennaceur, 2010; Heyser & Chemero, 2012; Lueptow, 2017). While a wide variety of  
64 protocols have been developed, the task involves two basic phases: a *familiarization* phase where the  
65 animal becomes acquainted with the objects and a *test* phase where the original familiar objects are  
66 changed (typically replaced by a different object, or moved to a novel location or context). During  
67 familiarization, the animal encodes the object's features, location, and context. During the test phase,  
68 due to the rodent's natural innate preference for novelty, it should spend more time investigating the  
69 modified object(s) compared to the unmodified object(s). Intact learning and memory are inferred  
70 based on increased investigation of the modified object(s) during the test phase, i.e., we infer that the  
71 animal recognizes the object as novel and thereby directs greater investigative behavior towards it  
72 (Heyser & Chemero, 2012; Leger et al., 2013; Lueptow, 2017).

73 These tasks are in widespread use, but there are many inconsistencies in the way they are conducted.  
74 Two major issues are the lack of consistency in the method of measuring object investigation and the  
75 lack of standardization of the objects used. The main methods of scoring object recognition tasks are  
76 video-based automated software and manual scoring. Current video-based automated systems can  
77 often be unreliable, lack temporal precision, and can be costly, whereas manual scoring is time-  
78 consuming, tedious, and subjective. A lack of standardization of the objects used in object recognition is  
79 also a concern in object recognition tasks. Examples of objects used during object recognition tasks  
80 include plastic toys, glass bottles, stacking squares, and metal cans. Object properties can differ across a  
81 large number of dimensions such as shape, texture, color, material, reflectivity, and size (Antunes &  
82 Biala, 2012; Benice & Raber, 2008; Bevins et al., 2002; Ennaceur, 2010; Heyser & Chemero, 2012; Leger  
83 et al., 2013; Lueptow, 2017). These properties strongly influence the investigation of the objects, and  
84 different affordances offered by objects can strongly bias the results (Ennaceur, 2010). Despite such

85 potential confounds, there has been little effort to develop a standardized approach for the selection of  
86 objects.

87 To resolve these issues, we sought to design and implement 3D printed objects for object recognition  
88 tasks, so that objects can be standardized and reproduced across labs. In addition, to promote a more  
89 standardized method for measuring object investigation, we developed a capacitive touch sensing  
90 approach to quantify investigation using the Arduino-based MPR121 capacitive touch sensor controller,  
91 making the objects themselves the sensors. We utilized a 3D printer and low-cost off-the-shelf  
92 components to aid in widespread adoption and cross-lab validation. The objects were tested in object  
93 recognition tasks and compared to manual scoring. Two options for the Capacitive Touch (CapTouch)  
94 system were created. CapTouch 1.0, which utilized conductive filament, and CapTouch 2.0, which  
95 utilized traditional filament combined with copper tape inside the object. CapTouch 2.0 was created  
96 after CapTouch 1.0 to provide an additional lower-cost method of creating objects that also provides  
97 more options for the printing material. We provide details about the materials and build instructions  
98 needed for both iterations as well as the validation from both sets of experiments.

## 99 **Materials and Methods**

### 100 *Components and Construction*

101 A basic diagram for the system is shown in Figure 1A. All components needed for the two CapTouch  
102 systems are listed in Tables 1 and 2, respectively. The estimated cost of making the CapTouch systems is  
103 about \$85.00 (this does not include Noldus products and the cost of constructing the chambers used to  
104 run the experiments). Detailed build instructions will be made available on the NIEHS Neurobehavioral  
105 Core Github. 3D models for the objects were designed using Blender 2.79, and Autodesk Netfab  
106 software was used to optimize the models prior to slicing. The final slicing and gcode generation were  
107 completed using PrusaSlicer and printed via Prusa i3 MK3s 3D printers, and this code will be available on  
108 the NIEHS Neurobehavioral Core Github.

109 The CapTouch 1.0 objects and their bases were printed with conductive filament to record interactions.  
110 The design allowed for easy removal and attachment of the object to the base via a twist-off design. The  
111 objects for CapTouch 1.0 were connected to the Bare Conductive Touch Board using solid core wire  
112 which was attached to a magnet that was inserted in the objects on one end and connected to the  
113 header pins on the touch board on the other end (Figure 2C).

114 The CapTouch 2.0 objects were printed in a non-conductive Flexfill filament and their associated bases  
115 in a PET-G filament. The objects were hollow to allow strips of copper foil tape to be placed inside the  
116 objects to serve as the capacitive sensor. The objects were connected to the bases by a simple peg-in-  
117 hole design. Solid core wire which was attached at one end to the base by copper tape was connected to  
118 the header pins on the touch board on the other end (Figure 3C). In both iterations, the CapTouch bases  
119 were connected to the arena floor using hot glue.

120 The touch board from Bare Conductive is an Arduino Leonardo based ATmega32U4 microcontroller  
121 board that runs at 16MHz from 5V with capacitive touch and MP3 decoder ICs. It uses a MPR121 chip  
122 that gives it twelve capacitive touch/proximity sensing electrodes (Bare Conductive, 2018). Capacitive  
123 touch/proximity sensing electrodes are devices that can detect the presence or absence of an object by  
124 utilizing a change in capacitance based on a change in the electrical field that is generated around the  
125 sensor. Capacitive touch/proximity sensors operate as a simple capacitor. The face of the object (the  
126 sensing face) is electrically connected to an internal oscillator circuit and the animal (the target) acts as  
127 the second plate of the capacitor which produces an electrostatic field. The external capacitance  
128 between the object and the animal forms part of the feedback capacitance of the oscillator circuit; when  
129 the animal approaches the sensor, the oscillations increase until the set threshold level is reached and  
130 activates an output. Capacitive touch/proximity sensors sensitivity can be adjusted which can change  
131 the operating distance to the target (Moermond, n.d.). The MPR121 uses an auto-calibration mechanism  
132 that detects background capacitance (which varies as a function of the size of the object used for the

133 sensor) and subtracts this to achieve an optimized baseline. The touch board can be programmed using  
134 the Arduino IDE and open source code was modified so that the touch/proximity threshold of the  
135 capacitive touch electrodes was set at 1 and the release threshold was at 2 (Bare Conductive, 2018;  
136 code available on the NIEHS Neurobehavioral Core Github). The Arduino IDE sketch was programmed to  
137 send a TTL pulse from a corresponding output pin when a touch electrode detected a touch. This TTL  
138 pulse was sent to a Noldus IO Box to be recorded by the analysis software, Ethovision. TTL signal could  
139 be read out via a variety of methods, however, we chose the Noldus IO Box as it would allow us to easily  
140 cross-validate with automated and manual scoring. Readout parameters included the total number of  
141 interactions, the amount of time of each interaction, and the total summed interaction time.

#### 142 *Animals*

143 Adult female and male C57BL/6 mice were obtained from Taconic Farms and were group-housed on a  
144 12-hour light/12-hour dark cycle. All experiments were performed during the dark phase. For the object  
145 preference test, four female and four male mice were used at three months of age. For the CapTouch  
146 1.0 experiments, eight female and eight male mice were used at four months of age. For the CapTouch  
147 2.0 experiments, four female and four male mice were used at three months of age. For the 24-hour  
148 retention interval experiments using CapTouch 1.0 objects, two sets of eight female and eight male mice  
149 were used at three months of age. All animal procedures were performed in accordance with the NIEHS  
150 animal care committee's regulations.

#### 151 *Procedure*

152 Before the novel object recognition task took place during the CapTouch 1.0 experiments, an object  
153 preference task was performed with naïve mice to assess the preference or lack of preference for one of  
154 the objects over the other. This is a critical step in the novel object recognition task because it ensures  
155 that there is not an innate preference for one of the objects, which could skew results obtained from the  
156 novel object recognition task. This task occurs in two phases: habituation and test. During the



157 habituation phase, the animal was placed in the middle of the open-field arena and was allowed to  
158 explore the open-field arena for ten minutes. During the test phase, one of each object was placed in  
159 the corners of the arena, the animal is introduced into the middle of the arena and allowed to explore  
160 the arena and objects for ten minutes. The objects were counterbalanced between the two arenas to  
161 account for side preference and they were placed directly in the corners against the walls. Duration of  
162 time each mouse spent investigating the objects was recorded and calculated to ensure there was no  
163 significant difference between the length of time spent investigating each object.

164 The novel object recognition task took place during three phases: habituation, familiarization, and test  
165 phase. In the habituation phase, the animal was allowed to explore a dimly lit (5-10 lux) open-field arena  
166 for ten minutes on two consecutive days (20 minutes total). During the familiarization phase, two  
167 capacitive touch objects were set up in the arena. The animal was placed in the center of the arena and  
168 was allowed to explore the arena and objects for ten minutes. After a retention interval of ten minutes,  
169 the test phase occurred. During the test phase, one of the objects was switched for a novel object and  
170 the second object stayed the same as during the familiarization phase. The animal was placed in the  
171 center of the arena and was allowed to explore the arena and objects for ten minutes.

172 A 24-hour retention interval was also tested in the novel object recognition task as used in earlier  
173 reports (Kwapis et al., 2019; Pereira et al., 2014; Tuscher et al. 2018; Vogel-Ciernia & Wood, 2014). In  
174 attempt to increase object investigation, the mice were exposed to different objects with the same 3D  
175 printed filament in their home cages before the familiarization and test phases for ten minutes while  
176 being handled on the day prior to the experiments. During the familiarization phase, mice were exposed  
177 to objects until they interacted for a cumulative 30 seconds or remained in the arena to explore the  
178 objects for 30 minutes. Mice that did not reach ten cumulative seconds of investigation during the  
179 familiarization phase were not included in the test phase. During the test phase, mice were allowed to  
180 explore the arena and objects for 30 minutes. All objects were placed in the corners of the arena directly

181 against the wall (Figure 1B shows a basic workflow of the novel object recognition experiment). The  
182 novel and familiar objects were counterbalanced across mice. Objects were thoroughly cleaned with  
183 70% ethanol between animals and allowed to dry and the arena was thoroughly cleaned with Windex  
184 between animals and allowed to dry. The Bare Conductive touch boards were reset before every trial to  
185 engage the auto-calibration mechanism of the MPR121 capacitive sensor to eliminate any subtle  
186 baseline capacitance changes that might have occurred during cleaning. Capacitive touch interaction  
187 was recorded via the Noldus IO box along with video using a Microsoft c930e Webcam at 800 x 600  
188 resolution. The camera was modified to detect only infra-red (IR) light by removing the IR cut filter and  
189 placing an IR-pass filter over the lens ([https://www.alcs.ch/logitech-c910-infrared-conversion-for-](https://www.alcs.ch/logitech-c910-infrared-conversion-for-nightvision.html)  
190 [nightvision.html](https://www.alcs.ch/logitech-c910-infrared-conversion-for-nightvision.html)), and IR light was used to illuminate the arena. This allowed for consistent quality video,  
191 even under dim lighting conditions.

#### 192 *Analysis*

193 The CapTouch system was compared to the manual scoring using the manual scoring feature of  
194 Ethovision. During manual scoring, the scorer considered investigation of the object to begin the video  
195 frame after the animal's nose orients towards the object within two centimeters of the object.  
196 Investigation ends the video frame that the animal's nose moves away from the object. For both  
197 CapTouch 1.0 and 2.0, two manual scorers were used. The scorers were aware of the overall goals of the  
198 project but were trained to use the parameters for what is considered investigation described above  
199 and scored both the familiarization and test phases of the experiments. The total investigation time for  
200 both objects was summed for each session and then utilized to calculate a Pearson correlation between  
201 manual and CapTouch scoring for all trials. Repeated measures ANOVA was used to analyze the novel  
202 object recognition task for the CapTouch 1.0 objects. Interaction with the familiar object and novel  
203 object during the test phase were compared by calculating a percent investigation for each object. For

204 the novel object: (novel investigation time/total time x 100) and for the familiar object: (familiar  
205 investigation time/(total time x 100)). .

#### 206 *Statistics*

207 Data were analyzed using IBM SPSS statistics. A Pearson correlation was used to examine the correlation  
208 between manual scoring and CapTouch scoring for all trials and significance was evaluated with a two-  
209 tailed t-test. Repeated measures ANOVA was used to analyze the novel object recognition task for  
210 CapTouch 1.0 objects and mean interaction time comparisons between manual and CapTouch.  
211 Statistical significance was considered at  $p < .05$ .

### 212 **Results**

#### 213 **Capacitive Touch 1.0**

##### 214 *Capacitive Sensing Validation*

215 To test the validity of the CapTouch 1.0 system, videos from the object preference task, the  
216 familiarization phase of the ten-minute delay novel object recognition task, and the test phase of the  
217 ten-minute delay novel object recognition task were compared to manual scoring. Utilizing the sum of  
218 object investigation across each session we found a high degree of correlation between manual scoring  
219 and the capacitive touch sensing (Pearson correlation,  $R^2=0.9216$ ,  $p < 0.0001$ , Figure 2H). Looking at  
220 average object interaction for both manual scoring and capacitive touch sensing, CapTouch 1.0 has a  
221 slightly lower mean interaction duration compared to manual scoring (21.11 sec +/- 4.68 SEM vs. 26.06  
222 sec +/- 4.73 SEM,  $F(1,51) = 13.816$ ,  $p = 0.001$ , partial eta squared = .213). When comparing the  
223 capacitive touch sensing against an additional scorer, a strong correlation was found (Pearson  
224 correlation,  $R^2=0.9167$ ,  $p < 0.0001$ , Extended Data Figure 2-1A), as well as when comparing the two  
225 manual scorers against each other (Pearson correlation,  $R^2=0.9950$ ,  $p < 0.0001$ , Extended Data Figure 2-  
226 1B). Figure 2I provides a representative 30 second example that compares the triggering of the  
227 capacitive sensing system against manual scoring. The time between object interactions, i.e. the inter-

228 interaction interval, was calculated and did not differ between manual and capacitive sensing (Extended  
229 Data Figure 2-2).

### 230 *Object Preference*

231 Figure 2D shows the percent investigation of the objects over ten-minute trials. It was shown that there  
232 was a lack of preference between the two objects ( $p=0.7776$ ). The spiral and sphere objects were then  
233 used in the novel object recognition task since an innate preference was not found.

### 234 *Novel Object Recognition*

235 Results from the ten-minute delay NOR task for the CapTouch 1.0 system show the mice had a  
236 significant preference for the novel object when compared to the familiar object during the test phase  
237 after a ten-minute retention interval ( $F(1,12) = 17.418$ ,  $p = 0.001$ , effect size: partial eta squared = 0.592,  
238 Figure 2E). This shows that learning occurred during the familiarization phase and short-term memory of  
239 the familiar object was intact during the test phase. Results from the 24-hour delay NOR task for the  
240 Capacitive Touch 1.0 system also shows that mice had a significant preference for the novel object  
241 compared to the familiar object during the test phase after a 24-hour retention interval ( $F(1,19) =$   
242  $10.615$ ,  $p = 0.004$ , partial eta squared = 0.358, Figure 2F). To further examine interaction preference for  
243 the novel object compared to the familiar object, a discrimination index ((Novel interaction time –  
244 Familiar Interaction Time)/Total Interaction Time)) was calculated for the ten-minute delay ( $p=0.0013$   
245 relative to a chance score of zero) and 24-hour delay ( $p=0.0165$  relative to a chance score of zero) NOR  
246 experiments (Figure 2G).

### 247 **Capacitive Touch 2.0**

#### 248 *Capacitive Sensing Validation*

249 To test the validity of the Capacitive Touch 2.0 system, the videos from the familiarization and test  
250 phase of the validation experiment were manually scored for object interaction and correlated against  
251 the TTL pulses the CapTouch 2.0 system picked up from object interaction. Using this methodology, we

252 observed a high degree of correlation between capacitive touch sensing and manual scoring (Pearson  
253 correlation,  $R^2=0.9767$ ,  $p\text{-value}<0.0001$ , Figure 3D). Similar to CapTouch 1.0, the average interaction for  
254 CapTouch 2.0 also had a slightly lower mean interaction compared to manual scoring (11.68 sec +/- 2.95  
255 vs. 13.21 sec +/- 3.37 SEM,  $F(1,18) = 15.607$ ,  $p = 0.027$ ). A strong correlation was found when comparing  
256 the capacitive touch sensing against an additional scorer (Pearson correlation,  $R^2=0.9316$ ,  $p <0.0001$ ,  
257 Extended Data Figure 3-1A), as well as when comparing the two manual scorers against each other  
258 (Pearson correlation,  $R^2=0.9642$ ,  $p <0.0001$ , Extended Data Figure 3-1B). Figure 3E provides a 30-second  
259 representative example that compares the triggering of the capacitive sensing system against manual  
260 scoring.

#### 261 **Discussion**

262 Here, we describe a novel approach to object recognition tasks utilizing 3D printed capacitive sensing  
263 which can be utilized to standardize the objects used and the method for scoring object investigation.  
264 Two iterations of the capacitive touch system were created and tested: CapTouch 1.0 and CapTouch 2.0.  
265 In our experiments for CapTouch 1.0, the objects were 3D printed with a conductive filament that  
266 allowed for the object itself to serve as a capacitive sensor. The objects were tested against each other  
267 and no preference was found between them, which is a critical validation step when choosing objects.  
268 Basic novel object recognition tests were performed that shows the system's accuracy compared to  
269 manual scoring and confirmed that mice were able to distinguish between the two objects, showing  
270 preference towards the novel object when introduced to it. The CapTouch 2.0 approach allows for the  
271 use of any 3D printed filament by making the objects hollow and coating the inside with copper tape to  
272 provide the object's conductivity. Both the CapTouch 1.0 and 2.0 approaches show a high positive  
273 correlation when compared against manual scoring from multiple scorers, indicating the CapTouch  
274 system is a reproducible and viable method regardless of the iteration used.

275 These experiments provide a proof-of-concept demonstration that capacitive touch sensing can be a  
276 reliable method for detecting investigation in object recognition tasks. We hope this research can pave  
277 the way for future studies to begin validating standardized object sets that can be used across labs and  
278 institutions. We validated an initial set of two object pairs, but our overall approach will allow for a  
279 concerted global effort to develop a standardized battery of objects that can be used to vary different  
280 dimensions of object properties, including size, color, and shape. The parameter space for this is quite  
281 large and will take a substantial effort to cross-validate across labs, but we feel that such an effort will  
282 be worthwhile for the field. Standardizing objects will help to reduce the current state of the field, which  
283 is characterized by a large variability in the types of objects that are used, as seen in the Supplemental  
284 Table. In addition, the more standardized and high-throughput method for detecting object  
285 investigation developed here will aid in this standardization effort by reducing the personnel time  
286 required to obtain accurate data. Finally, this system can be implemented at a relatively low cost as it  
287 uses inexpensive, off-the-shelf components to easily allow labs to conduct their own studies and  
288 potentially participate in cross-validation studies.

289 The capacitive touch system has the possibility to be versatile and modified to the user's needs. There is  
290 noteworthy opportunity to create different object designs and choose different colors using 3D printing.  
291 Also, the sensitivity of the capacitive touch sensing can be adjusted at both the hardware and software  
292 levels, allowing the system to be more sensitive or less sensitive to interaction and scaled to work with a  
293 range of rodent sizes. This capability allows for this system to easily transition between different rodent  
294 models in object-recognition assays.

#### 295 **System limitations**

296 Despite our convincing proof-of-principle demonstration, capacitive sensing does have some limitations  
297 that will be addressed in future iterations. First, it detects slightly less interaction than manual scoring  
298 does due a greater requirement for direct physical contact. This is clearly still sufficient for conducting

299 object recognition experiments, as we demonstrate here, but could miss brief exploratory interactions  
300 that may be particularly relevant when combining with neural recordings. The second limitation, found  
301 only in the CapTouch 1.0 model, is the use of conductive filament. This type of filament is more  
302 expensive and has limited color options compared to other types of filament. The limitation in filament  
303 colors restricts the range of available objects, so we created CapTouch 2.0 as an alternative.

304 The third limitation with the CapTouch 1.0 model was that once the mice became comfortable around  
305 the objects, they began climbing and sitting on them. Climbing in object recognition tasks is often  
306 debated on whether it should be considered object investigation. Climbing has been associated with  
307 significantly longer exploration, slower habituation to objects, and higher discrimination in objects  
308 (Heyser & Chemero, 2012). This alters the object investigation data because the mice are no longer  
309 investigating the objects but instead using the objects as a pedestal to gain a different vantage point of  
310 the area. To combat this limitation, CapTouch 2.0 objects were created with pointed tops to deter the  
311 mice from climbing and sitting on the objects, which may have contributed to the overall reduced object  
312 investigation between the two methods.

313 Perhaps the most serious limitation could be the 3D printing filament itself. Our experience suggests  
314 that the level of object investigation may be less than what we would typically expect. In hopes to  
315 increase object investigation, before running the 24-hour retention interval novel object recognition  
316 experiment, the mice were exposed to objects with the same 3D printed material as the objects that  
317 were being used in the familiarization and test phases while being handled. It seems that this exposure  
318 helped increase investigation with some mice but not all. The variability seen within a cohort of mice in  
319 the different levels of investigation could be due to the sensitivity of the mice to volatile organic  
320 compounds that off-gas from the objects. A recent report suggests this could be the case (Tropea et. al,  
321 2019), however future work needs to be done to determine the extent to which this is, in fact, a

322 problem and what methods can be employed to mitigate it (providing sufficient time to off-gas any  
323 aversive volatile compounds, etc.).

#### 324 **Future directions**

325 The most immediate future directions focus on addressing the limitations presented above. First, we are  
326 currently investigating ways to streamline the capacitive touch system setup to make it more reliable  
327 and easier to implement. Similar to other DIY projects, we plan to support further development,  
328 implementation, and standardization experiments by hosting a web-based forum to collaboratively track  
329 progress. Our initial major goals are to make the overall system more robust to facilitate the ease of set  
330 up and take-down of the components. We plan to develop a standalone data readout system that is  
331 low-cost and does not require third-party hardware, similar to (Ardesch et al., 2017). The goal would be  
332 to read-out the raw analog capacitance values so that touch and release thresholds could be tweaked  
333 off-line as needed rather than hard coded into the Arduino code. This would give us greater flexibility to  
334 titrate the sensitivity to ensure all interactions are detected. Our long term and most ambitious goal is  
335 to spur an effort across multiple labs to develop and standardize object sets, similar to stimulus sets in  
336 human psychology (Olszanowski et al., 2014)

#### 337 **Conclusion**

338 The capacitive touch system presented here provides investigators with a low-cost and easily  
339 reproducible system to score object investigation in rodents. The 3D printed object capabilities and the  
340 open source availability of this system could be used to standardize objects used in object recognition  
341 assays across labs. Widespread use of standardized objects and methods for measuring investigation  
342 would revolutionize the use of object recognition tasks. This would ultimately lead to a better  
343 understanding of the basic mechanisms of learning and memory and substantially improve animal  
344 models of neurodegenerative and neuropsychiatric disorders overall.

345



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391

392 **Component List for Capacitive Touch 1.0**

Component	Quantity	Supplier	Part Number	Price
Touch Board	1	Bare Conductive	NA	\$49.91 (List price: \$62.39)
Electric Paint 10ml	1	Bare Conductive	NA	\$11.04
Perma-Proto Half-sized Breadboard PCB- Single	1	Adafruit	1609	\$4.50
Proto- Pasta Conductive PLA- 1.75mm (.5kg)	1	MatterHackers	MUW33A27	\$49.99 (List price: \$56.00)
RJ45 8- Pin Connector	1 per touch board	SparkFun	PRT-00643	\$1.50
Short Headers Kit for Feather- 12- pin + 16- pin Female Headers	2 per touch board	Adafruit	2940	\$1.50
Solid- Core Wire Spool- 25ft- 22AWG	1	Adafruit	290	\$2.95
Magnet- 1/2" dia. X 1/10" thick	1 per object	K&J Magnetics	D8H1	\$0.83
Diffused 5mm LED	1	Adafruit	299	\$4.00

(25 pack)				
USB- IO Box	1	Noldus	NA	\$1535.00
Ethernet Cable	1 per touch board	Adafruit	994	\$2.75
USB cable- USB A to Micro- B- 3 foot long	1 per touch board	Adafruit	592	\$2.95
Resistor- 10K ohm -Pack of 25	2 per touch board	Adafruit	2784	\$0.75

393 **Table 1.** CapTouch 1.0 list of build components needed for the system. The component name, number  
394 needed, supplier, part number, and price are provided.

395

396

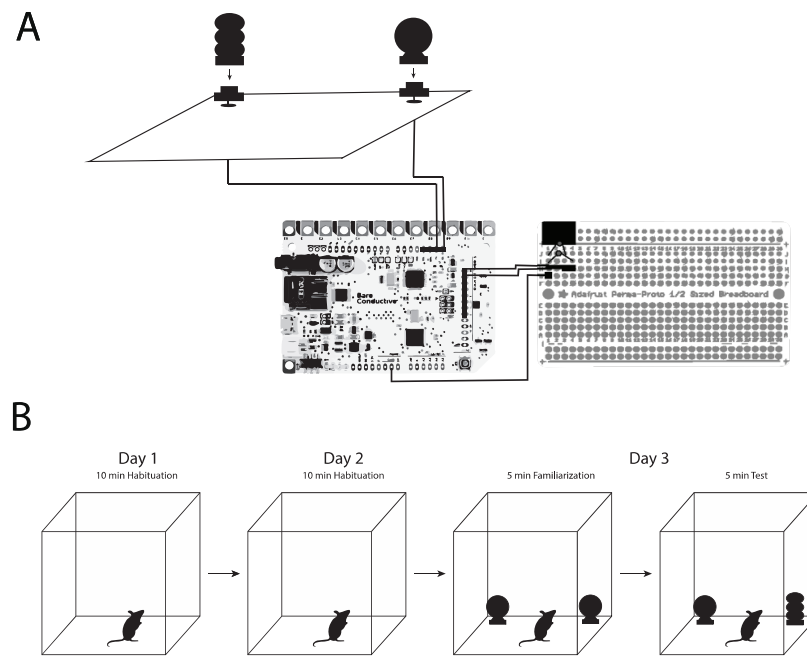
397 **Component List for Capacitive Touch 2.0**

Component	Quantity	Supplier	Part Number	Price
Flexfill 98A Powder Beige filament 500g	1	Prusa Research	FLM-FLX-175- PBG-98A	\$33.99
Flexfill 98A Luminous Green filament 500g	1	Prusa Research	FLM-FLX-175- GRN-98A	\$33.99
PETG Prusa Orange filament 1kg	1	Prusa Research	PRM-PETG-PRO- 1000	\$29.99
Touch Board	1	Bare Conductive	NA	\$49.91 (List price: \$62.39)
Solid- Core Wire Spool- 25ft- 22AWG	1	Adafruit	290	\$2.95
Perma-Proto Half- sized Breadboard PCB- Single	1	Adafruit	1609	\$4.50
RJ45 8- Pin Connector	1 per touch board	SparkFun	PRT-00643	\$1.50
Short Headers Kit for Feather- 12- pin	2 per touch board	Adafruit	2940	\$1.50

+ 16- pin Female Headers				
Diffused 5mm LED (25 pack)	1	Adafruit	299	\$4.00
USB- IO Box	1	Noldus	NA	\$1535.00
Ethernet Cable	1 per touch board	Adafruit	994	\$2.75
USB cable- USB A to Micro- B- 3 foot long	1 per touch board	Adafruit	592	\$2.95
Resistor- 10K ohm	2 per touch board	Adafruit	2784	\$0.75
Copper Foil Tape with Conductive Adhesive- 25mm x 15 meter roll	1	Adafruit	1127	\$19.95

398 **Table 2.** CapTouch 2.0 list of build components needed for the system. The component name, number  
399 needed, supplier, part number, and price are provided.  
400

401 **Figure 1.**



402

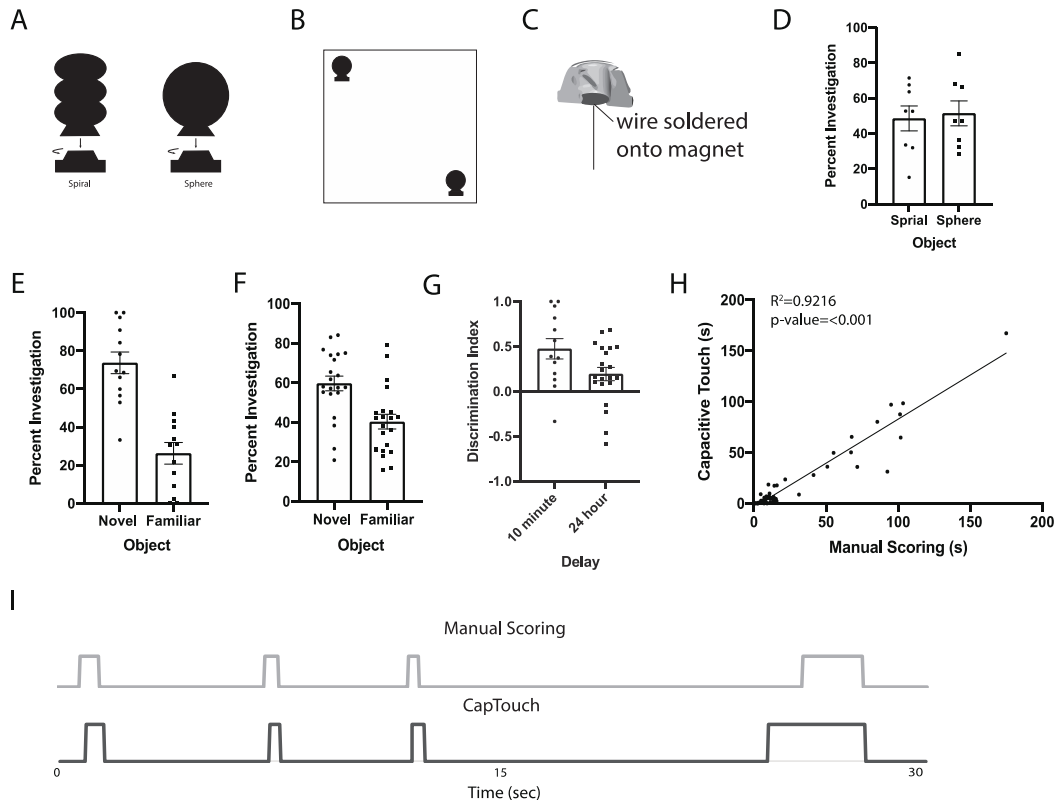
403 **Figure 1. CapTouch Overview.** A, Diagram of basic CapTouch system setup with Bare Conductive

404 capacitive sensing board and breadboard. B, Basic novel object recognition workflow.

405



406 **Figure 2.**

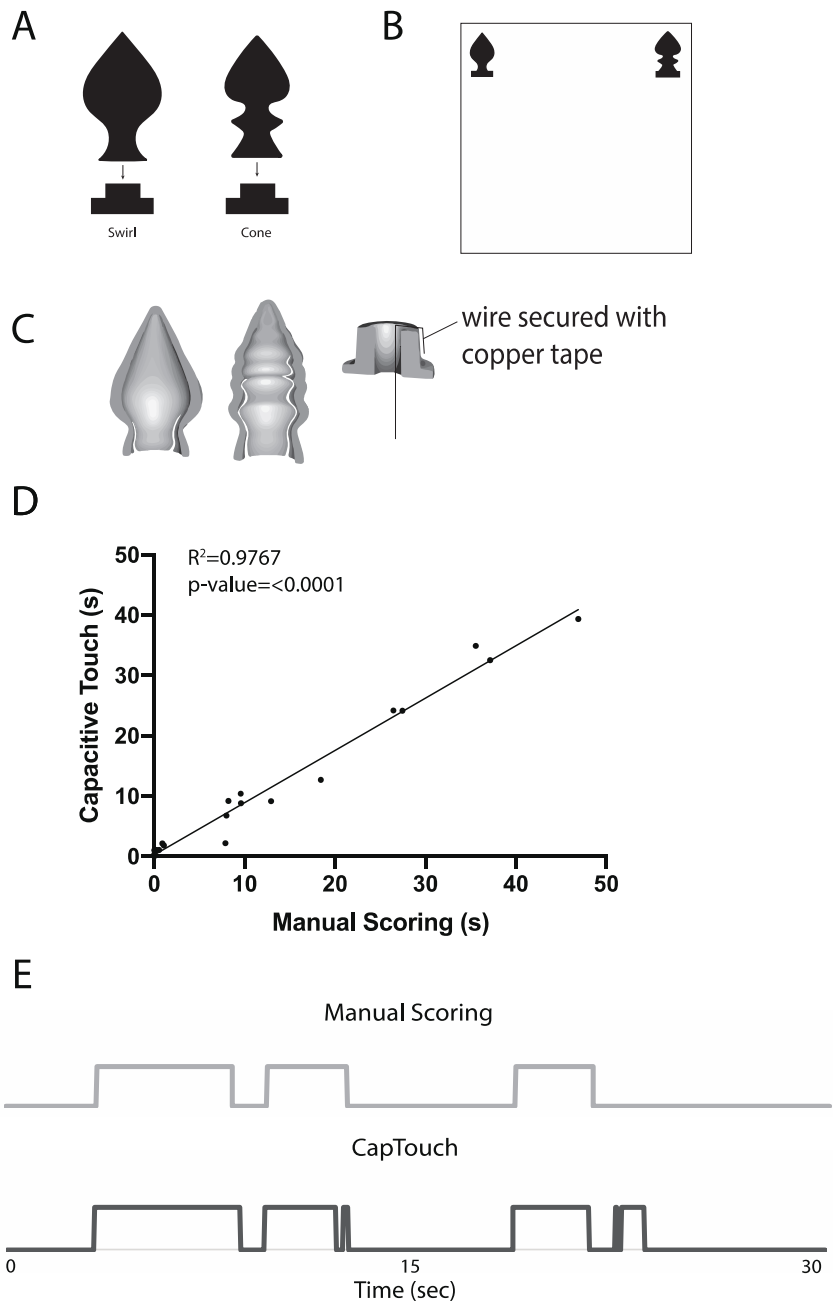


407

408 **Figure 2. CapTouch 1.0 NOR results and validation.** A, Diagram of CapTouch 1.0 objects spiral (right)  
 409 and sphere (left). B, Object placement during NOR experiment. C, Cross sectional view and diagram of  
 410 wire attachment to the base. D, Percent investigation between the spiral and sphere object during the  
 411 object preference test. (+/- SEM) E, Percent investigation between the novel and familiar object with a  
 412 ten-minute delay during the NOR experiment. Two mice were excluded from the novel object  
 413 recognition results due to CapTouch sensing malfunction. (+/- SEM) F, Percent investigation between  
 414 the novel and familiar object with a 24-hour delay during the NOR experiment. (+/- SEM) G,  
 415 Discrimination index comparing investigation between the novel and familiar objects for the ten-minute

416 delay ( $p=0.0013$  relative to chance) and the 24-hour delay ( $p=0.0165$  relative to chance) NOR  
417 experiments. H, Correlation of object investigation duration (sec) during the object preference test and  
418 both the familiarization and test days of the NOR test between capacitive touch sensing and manual  
419 scoring ( $R^2=0.9162$ ,  $p\text{-value}<0.001$ , see Extended Data Figure 2-1 for further validation with an additional  
420 manual scorer). I, 30 second example of capacitive touch triggering compared to manual scoring. For  
421 additional validation analysis see Extended Data Figure 2-2.  
422

423 Figure 3.



424

425 **Figure 3. CapTouch 2.0 validation.** A, Diagram of CapTouch 2.0 objects swirl (left) and cone (right). B,  
426 Object placement during system validation experiment. C, Cross sectional view of hollow objects and  
427 diagram of wire attached to base. D, Correlation of object investigation duration (sec) for both  
428 familiarization and test days between capacitive touch sensing and manual scoring ( $R^2=0.9767$ , p-  
429 value= $<0.0001$ , see Extended Data Figure 3-1 for validation of an additional manual scorer). Because of  
430 low interaction from some of the mice during the CapTouch 2.0 validation experiment, two highly  
431 interactive mice were run through an additional trial during the familiarization and test phases to  
432 acquire more interaction data. One object during a trial had a CapTouch sensing malfunction and was  
433 not included in the validation correlation. E, 30 second example of capacitive touch triggering compared  
434 to manual scoring.

435

#### 436 **Extended Data Figure Legends**

437 **Extended Data Figure 2-1.** A, Correlation of object investigation duration (sec) between capacitive touch  
438 sensing and an additional manual scorer ( $R^2=0.9167$ , p-value= $<0.0001$ ). B, Comparison of manual scorers  
439 correlated against each other ( $R^2=0.9950$ , p-value= $<0.0001$ ).

440

441 **Extended Data Figure 2-2.** A, Histogram of time between object interactions (inter-interaction interval)  
442 for capacitive sensing (mean= $13.9\pm 2.7s$ ) and manual scoring (mean= $17.3\pm 1.7s$ ) for the CapTouch 1.0  
443 experiments (p $>0.05$ , Moody test). B, Violin plot of inter-interaction intervals for capacitive sensing and  
444 manual scoring for the CapTouch 1.0 experiments.

445

446 **Extended Data Figure 3-1.** A, Correlation of object investigation duration (sec) between capacitive touch  
447 sensing and an additional manual scorer ( $R^2=0.9313$ , p-value= $<0.0001$ ). B, Correlation of manual scorers  
448 against each other ( $R^2=0.9642$ , p-value= $<0.0001$ ).

