1 Hierarchical architecture of dopaminergic circuits enables second-order conditioning

- 2 in Drosophila
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17 Abstract

18 Dopaminergic neurons with distinct projection patterns and physiological properties 19 compose memory subsystems in a brain. However, it is poorly understood whether or 20 how they interact during complex learning. Here, we identify a feedforward circuit formed 21 between dopamine subsystems and show that it is essential for second-order 22 conditioning, an ethologically important form of higher-order associative learning. The 23 Drosophila mushroom body comprises a series of dopaminergic compartments, each of 24 which exhibits distinct memory dynamics. We find that a slow and stable memory 25 compartment can serve as an effective "teacher" by instructing other faster and transient 26 memory compartments via a single key interneuron, which we identify by connectome 27 analysis and neurotransmitter prediction. This excitatory interneuron acquires enhanced 28 response to reward-predicting odor after first-order conditioning and, upon activation, 29 evokes dopamine release in the "student" compartments. These hierarchical 30 connections between dopamine subsystems explain distinct properties of first- and 31 second-order memory long known by behavioral psychologists.

32

33 Introduction

34 Knowledge about order and regularities in environments is crucial for animal survival. 35 Although direct temporal correlation between stimuli and rewards is a primary drive for 36 associative learning, animals are also capable of learning indirect relations between 37 stimuli and rewards in many real-life situations. For example, bumble bees, who have 38 prior foraging experience with other bees, can learn to visit a flower of a particular color 39 without tasting nectar just by watching other bees sitting on flowers of that color 40 (Avarguès-Weber and Chittka, 2014; Worden and Papaj, 2005a). In the case of humans, 41 some TV commercials can be considered as conditioning of consumers to associate 42 items with the positive valence that has been already associated with popular cartoon

43 characters. In both cases, learning depends on the valence of stimuli (i.e. sight of other 44 bees or cartoon characters) that is acquired through prior experience. Although such 45 higher-order associative learning is widely observed across species and ethologically 46 important, its circuit mechanisms are poorly understood compared to those of simpler 47 forms of associative learning.

48 Second-order conditioning is a major form of higher-order associative learning. In 49 this learning paradigm, an initially neutral stimulus is paired with reward or punishment; 50 that stimulus, which is now predictive of reward/punishment, then serves as an effective 51 reinforcer when learning about a new stimulus. Since Pavlov's classic experiment with 52 dogs (Pavlov, 1927), second-order conditioning has been demonstrated in various 53 vertebrate and invertebrate models (Bitterman et al., 1983; Brembs and Heisenberg, 54 2001; Hawkins et al., 1998; Holland and Rescorla, 1975; Mizunami et al., 2009; Rizley 55 and Rescorla, 1972; Sisk, 1976; Tabone and de Belle, 2011; Takeda, 1961). 56 Furthermore, second-order conditioning is thought to extend the applicability of 57 Pavlovian conditioning as an account of behaviors including observational learning 58 (Avarguès-Weber and Chittka, 2014; Worden and Papaj, 2005b). Additionally, second-59 order conditioning has also served as a historically important tool for behavioral 60 psychologists to study associative learning by giving them ample options to use virtually 61 any stimulus as a reinforcer (Rescorla, 1980).

One prominent feature that characterizes second-order memory is its transiency, as originally noted by Pavlov and confirmed by other studies using various animal models (Herendeen and Chris Anderson, 1968; Stout et al., 2004; Yin et al., 1994). That is, the effectiveness of second-order conditioning usually reaches an asymptote after a small number of trials and begins to decline with further training (Gewirtz and Davis, 2000; Pavlov, 1927). This decline may be related to the fact that reward is constantly omitted during second-order conditioning. Another important feature of second-order

69 conditioning recognized by behavioral psychologists is that it does not form a tight 70 association between the stimulus and the specific response elicited by the reinforcer, 71 which is typically observed in first-order conditioning (Gewirtz and Davis, 2000; Pavlov, 72 1927). In other words, second-order learning seems to be based on general valence, 73 rather than specific features, of reinforcers. These differences between first- and 74 second-order memories raise important mechanistic questions: What is the circuit origin 75 of those different memory features? Are they different because those two memories are 76 stored in separate circuits that support distinct types of memories? If so, how do the two 77 circuits interact when one memory instructs the other? Answering these questions 78 requires precise mapping of second-order memory circuits.

79 In rodents, basolateral amygdala and dopaminergic neurons (DANs) play critical 80 roles in second-order learning (Gewirtz and Davis, 1997; Maes et al., 2020). After first-81 order association. DANs in the ventral tegmental area acquire enhanced responses at 82 the onset of the cue that predicts upcoming reward after conditioning (Schultz, 1998). A 83 recent study used optogenetic silencing to demonstrate that such cue-evoked dopamine 84 transients are essential for second-order conditioning (Maes et al., 2020). Whereas 85 DANs consist of functionally diverse populations of neurons, each of which contributing 86 to distinct types of learning (Roeper, 2013; Watabe-Uchida and Uchida, 2018), how 87 these different DAN subtypes interact during second-order conditioning is completely 88 unstudied.

The *Drosophila* mushroom body (MB), a dopamine-rich center for associative learning in insect brains, provides a tractable system to study the interaction between heterogeneous dopamine subsystems. *Drosophila* can perform second-order learning using olfactory or visual cues with punishment (Brembs and Heisenberg, 2001; Tabone and de Belle, 2011), although the underlying circuit mechanisms have not been examined. Decades of studies have revealed the anatomical and functional architecture

95 of the MB circuit (Figure 1A). Along the parallel axonal fibers of Kenyon cells (KCs), 96 DANs and MB output neurons (MBONs) form 16 matched compartments (Aso et al., 97 2014; Li et al., 2020; Tanaka et al., 2008), which serve as units of associative learning. 98 Reward and punishment activate distinct subsets of 20 types of DANs (Berry et al., 99 2015; Burke et al., 2012; Kirkhart and Scott, 2015; Lewis et al., 2015; Lin et al., 2014; Liu 100 et al., 2012; Riemensperger et al., 2005; Siju et al., 2020). Individual DANs write and 101 update memories in each compartment with cell-type-specific dynamics by modulating 102 synaptic connection between KCs and MBONs (Aso et al., 2019, 2012; Aso and Rubin, 103 2016; Hige et al., 2015; Huetteroth et al., 2015; Owald et al., 2015; Perisse et al., 2016; 104 Vrontou et al., 2021; Yamagata et al., 2015). Outside the MB, MBON axons project to 105 regions where DAN dendrites arborize; this provides an anatomical pathway for 106 feedback of memory-based information onto DANs, a potential substrate for higher-order 107 conditioning. Indeed, early studies showed that DANs in the MB dynamically change 108 odor responses after olfactory conditioning (Riemensperger et al., 2005). Furthermore, 109 the recently completed EM connectome (Scheffer et al., 2020) revealed the full wiring 110 diagram of the MB, including intricate connections from MBONs to the DANs. In both 111 larval and adult Drosophila, large fractions of synaptic inputs to the MB's DANs originate 112 from the MB itself (Eschbach et al., 2020; Li et al., 2020). Thus it is plausible that 113 induction of synaptic plasticity in one compartment, in turn, affects how a learned 114 stimulus activates DANs and becomes a secondary reinforcer. However, understanding 115 the flow of information across compartments that underlies second-order conditioning is 116 a challenging task, given that thousands of neurons are connected with DANs and 117 MBONs.

Here, by exploiting connectomic data, we identify a key circuit that underlies second-order conditioning. We first establish a protocol for robust olfactory second-order conditioning with sugar reward. In contrast to stable odor-sugar first-order memory,

121 second-order memory decayed within a day and was highly susceptible to extinction. We next show that memory in α_1 , the compartment responsible for long-lasting appetitive 122 123 memory (Ichinose et al., 2015; Yamagata et al., 2015), is most potent to promote 124 second-order memory. The second-order memory instructed by $\alpha 1$ was transient during 125 the training phase and extinction trials. Subsequent EM connectome and functional 126 analysis identify a prominent cholinergic interneuron SMP108 that 1) forms an excitatory 127 pathway from MBON-α1 to DANs in other compartments, 2) acquires an enhanced 128 response to the reward-predicting odor, 3) can promote release of dopamine in multiple 129 compartments, 4) is required for second-order conditioning, and 5) induces memory with 130 fast and transient dynamics. Our study reveals in unprecedented detail circuit 131 mechanisms of second-order conditioning. These mechanisms can explain the different 132 properties of first- and second-order memories. They also provide a concrete example of 133 how hierarchical interaction between dopamine subsystems contributes to a complex 134 form of learning.

135

136 Results

137 Olfactory second-order conditioning following the odor-sugar association

138 As a prerequisite for mapping the underlying neuronal circuits and detailed 139 characterization of memory properties, we established a robust protocol for appetitive 140 second-order conditioning using a circular olfactory arena (Figure 1B and Figure 1-figure 141 supplement 1; see Methods for our rationale for the selection of odors and other 142 parameters). Flies were first trained to associate stimulus one (S1) odor with sugar and 143 consolidated that memory for one day (Figure 1C). During second-order conditioning, 20 144 seconds of one S2 odor (S2+) was immediately followed by 10 seconds of the S1 odor, 145 whereas another S2 odor (S2-) was presented alone. After five training sessions, flies 146 increased their preference to the S2+ odor over the S2- odor when first-order 147 conditioning was long enough (i.e. 5 min; Figure 1D). This preference for the S2+ odor
148 was not due to sensory preconditioning, another form of higher-order conditioning in
149 which S2-S1 pairing was done *before* pairing S1-sugar (Figure 1E), although unimodal
150 sensory preconditioning has been reported in aversive olfactory learning in *Drosophila*151 (Martinez-Cervantes et al., 2022).

152 First-order memory and its derived second-order memory exhibited marked 153 differences in dynamics of formation and update. Second-order memory after odor-sugar 154 conditioning did not last for one day and was susceptible to extinction (Figures 1F and 155 G). With optogenetic stimulation of sugar sensory neurons, the first-order memory 156 steadily increased during nine training sessions, whereas second-order memory peaked 157 at the third training and declined subsequently (Figure 1H). This transiency of learning 158 was not observed when activation of sugar sensory neurons was not omitted during 159 second-order conditioning (Figure 1H). Learning of association between S2+ odor and 160 activation of sugar sensory neurons was compromised when S2+ is preceded by S1 161 which predicts the occurrence of reward (Figure 11). These results indicate that the 162 transient and unstable nature of second-order memory observed across animal phyla also applies to Drosophila, and the temporal order of the stimuli is crucial for second-163 164 order conditioning as in first-order conditioning.

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166 Identification of MB compartments that instruct second-order conditioning

To identify the circuit elements that might be particularly important for secondorder conditioning, we examined whether first-order memory in certain MB compartments is more potent for instructing second-order conditioning than others. For this purpose, we substituted sugar with optogenetic activation of DANs to induce memory in a defined set of compartments (Figure 2A). Flies were first trained by pairing the S1 odor with optogenetic activation of specific DANs with CsChrimson

173 (see below for measurement of dopamine release). Then, the compartment-specific 174 memory of the S1 odor was tested for its power as a reinforcer in second-order 175 conditioning. Among four sets of DAN cell types that can induce first-order appetitive memory (Figure 2A), two sets — PAM- α 1 and a combination of PAM-v5 and β '2a — 176 177 could induce significant second-order memory compared to the genetic control (Figure 178 2B). Similar to first-order conditioning, stimulus timing was an important factor for 179 successful second-order conditioning (i.e. S2+ must precede S1; Figure 2C). PAM-α1 is 180 known to be essential for learning nutritional value and is required for long-term 181 appetitive memory (Yamagata et al., 2015), whereas memory induced by combinatorial 182 activation of PAM-y5 and PAM- β '2a is short-lasting (Aso and Rubin, 2016). As expected 183 from those different stabilities of the first-order memory, memory in PAM-α1 but not 184 PAM- $\gamma 5/\beta' 2a$ could instruct second-order conditioning one day after the first-order 185 conditioning (Figure 2B). Consistent with the outcome of this optogenetic experiment, 186 blocking of synaptic transmission from PAM-a1 DANs with Tetanus Toxin (TNT) light 187 chain abolished both S1 preference and second-order memory when assayed one day after odor-sugar conditioning (Figure 2D). In contrast, blocking PAM cluster DANs in the 188 189 y4, y5, β '2a with TNT impaired the second-order conditioning without affecting S1 190 preference (Figure 2D). The second-order memory derived from the first-order memory 191 in the $\alpha 1$ compartment exhibited the transient learning curve (Figure 2E-F) and 192 susceptibility to extinction, recapitulating observations after odor-sugar conditioning 193 (Figure 1F-H). Thus, these results suggest $\alpha 1$ as the primary candidate compartment to 194 store the first-order memory that instructs second-order conditioning. The first-order 195 memory in the $\gamma 5/\beta' 2a$ compartments may have a supplemental contribution to second-196 order conditioning, especially shortly after the first-order conditioning.

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199 Memory in α1 can instruct secondary plasticity across compartments

200 Memories and plasticity induced in different MB compartments differ in their properties 201 including retention, induction threshold and resistance to extinction (Aso et al., 2012; 202 Aso and Rubin, 2016; Hige et al., 2015; Huetteroth et al., 2015; Jacob and Waddell, 203 2020; Lin et al., 2014; Pai et al., 2013; Plaçais et al., 2013; Vrontou et al., 2021; 204 Yamagata et al., 2015). The markedly distinct memory dynamics between first- and 205 second-order memories noted above prompted us to hypothesize that those memories 206 are formed in different MB compartments. For aversive memory, transient inactivation of 207 MBON-y1pedc (a.k.a MB-MVP2), which mimics the effect of synaptic depression caused 208 by aversive learning, can serve as reinforcement (König et al., 2019; Ueoka et al., 2017). 209 Thus, if our hypothesis is correct, and if the $\alpha 1$ compartment indeed is potent for 210 instructing second-order conditioning, then local induction of synaptic plasticity in $\alpha 1$ 211 should drive secondary plasticity in other compartments during second-order 212 conditioning. Since PAM- γ 5 and β '2a can induce robust appetitive memory that is short-213 lasting and susceptible to extinction (Figure 2A) (Aso and Rubin, 2016), we reasoned 214 that second-order memory may involve compartments targeted by these DANs. To test 215 this idea, we first generated a split-LexA driver to express ChrimsonR selectively in 216 PAM- α 1 (Figure 3 -figure supplement 1). We then labeled either MBON- α 1 or MBON-217 $\gamma 5\beta' 2a$ by split-GAL4 lines to make whole-cell recordings from them (Figures 3A and 218 Figure 3 -figure supplement 2A). In MBON- α 1, we found that pairing an odor and DAN 219 activation leads to reduced spiking responses to that odor as in other MB compartments 220 examined in previous studies (Figure 3-figure supplement 2) (Berry et al., 2018; Handler 221 et al., 2019; Hige et al., 2015; Owald et al., 2015; Owald and Waddell, 2015; Séjourné et 222 al., 2011; Vrontou et al., 2021). MBON- γ 5 β '2a, on the other hand, did not elicit action 223 potentials that are readily distinguishable from synaptic potentials in response to odor 224 presentation or current injection (Figure 3 -figure supplement 3). We therefore focused on subthreshold responses. After a single round of second-order conditioning, MBON- $\gamma 5\beta'2a$ showed reduced responses to the S2+ odor, while responses to S2- did not change even after five repetitions of conditioning (Figures 3B and C). Repeated presentation of S2 odors without S1 did not cause a reduction of odor responses (Figures 3D and E). These results indicate that the α 1 compartment can instruct secondorder conditioning in the $\gamma 5/\beta'2a$ and potentially other compartments.

231

232 Candidate interneurons to mediate instruction signals for second-order233 conditioning

234 We next set out to identify the neuronal pathway responsible for the induction of second-235 order plasticity. MBON- $\alpha 1$ is the sole output pathway from the $\alpha 1$ compartment and is, 236 like other reward memory compartment MBONs, glutamatergic. Glutamate functions as 237 an inhibitory neurotransmitter with glutamate-gated-chloride channel (Liu and Wilson, 238 2013), although activity of glutamatergic MBONs can have a net excitatory effect on 239 DANs via other receptors or indirect pathways (Cohn et al., 2015; Ichinose et al., 2015; 240 Karuppudurai et al., 2014; Zhao et al., 2018). Upon induction of plasticity, MBON- α 1's 241 responses to learned odor will be depressed (Figure 3-figure supplement 2). Therefore, 242 if glutamate is inhibitory, the downstream circuits of the MBON- α 1 could gain an 243 enhanced response to a learned odor as an outcome of reduced inhibition, could feed an 244 excitatory drive to DANs for second-order conditioning, provided that there are such 245 connections. However, $\alpha 1$ appears to be an exceptionally isolated compartment. MBON-246 α 1 is the only MBON that does not send direct output to DANs innervating other 247 compartments; rather it only directly connects with the DANs that innervate the same 248 compartment, PAM- α 1 (Figure 4-figure supplement 1A)(Li et al., 2020). Similarly, 249 MBON- α 1 shows very limited connections to DANs innervating other compartments that 250 are mediated by a single interneuron (one-hop pathways; Li et al., 2020; Figure 4-figure supplement 1B). This led us to explore pathways with two interneurons between MBONα1 and DANs (two-hop pathways).

253 To explore pathways with interneurons between MBON- α 1 and DANs, we 254 queried the hemibrain EM connectome database (Li et al., 2020; Scheffer et al., 2020). 255 We then used a pre-trained machine learning algorithm to predict the most likely 256 neurotransmitters used by the connected neurons (Eckstein et al., 2020). Supplementary 257 File 1 summarizes the full connection matrix, neurotransmitter predictions for the 396 258 major interneuron cell types with at least 100 total synapses with MBONs and DANs. In 259 this way (see Methods for detail), we identified prominent cholinergic two-hop pathways 260 from MBON- α 1 to multiple reward-DANs including PAM-y5, y4, β '2a, β '2m, β '2p that 261 were mediated by the interneurons SMP353/354 and SMP108 (Figure 4A; Figure 4-262 supplement 2). The SMP108 is an outstanding cell type in many features. Among all 263 cholinergic neurons, SMP108 has the highest number of connections with reward DANs 264 (Figure 4-figure supplement 3). SMP108 also synapses onto all three cholinergic 265 interneurons (SMP177, LHPV5e1, LHPV10d1) in the second layer of the two-hop 266 pathways, providing additional excitatory drive to PAM DANs (Figure 4B). Intriguingly, 267 SMP108 also appeared as an outstanding cell type to receive direct inputs from MBON-268 $y5\beta'2a$ and output to DANs (Figure 4C). As discussed above, we identified the $y5/\beta'2a$ 269 as additional compartments that, like $\alpha 1$, can instruct second-order memory. Taken 270 together, among other candidate cell types such as CRE011 and LHPD5d1 (Figure 4C), 271 the circuit centered at SMP108 appears to be a prominent candidate that converts first-272 order plasticity in both $\alpha 1$ and $\gamma 5\beta' 2a$ compartments to excitatory drive to DANs.

Identification of SMP108 and its associated circuits allowed us to construct a few
testable hypotheses regarding the circuit mechanisms of second-order conditioning. First,
SMP108's response to the reward-predicting S1 odor should be potentiated after firstorder conditioning. Second, activation of SMP108 should trigger dopamine release in the

277 MB compartments involved in appetitive memory. Third, the output of SMP108 should be 278 required for second-order memory. Fourth, memory induced by the SMP108 pathway 279 should recapitulate the transient and unstable nature of second-order memory. To 280 experimentally test those hypotheses, we generated split-GAL4 drivers for SMP108 281 (SS67221 and SS45234; Figures 4D-F). Using these drivers, we confirmed that axonal 282 terminals of SMP108 are immunoreactive to choline acetyltransferase (Figure 4E), which 283 is consistent with the fact that 2,416 out of 2,753 presynaptic sites of SMP108 are 284 predicted to be cholinergic in the hemibrain data (Supplementary File 1).

285

286 SMP108 acquires enhanced response to reward-predicting odor

287 First, we examined the change in SMP108's odor responses after pairing of an odor and 288 optogenetic activation of PAM-cluster DANs, which can induce appetitive memory. As 289 expected from the converging inputs from multiple lateral horn cell types (Supplementary 290 File 1), SMP108 showed robust spiking responses to odors. After pairing, responses to 291 the paired odor were selectively potentiated (Figure 5). Furthermore, reversal pairing de-292 potentiated the previously paired odor. Thus, SMP108 is capable of acquiring enhanced 293 responses to S1 after first-order conditioning and flexibly tracking updates of odor-294 reward associations.

295

296 SMP108 evoked dopamine release in appetitive memory compartments

Next, we directly measured the pattern of dopamine release evoked by optogenetic activation of SMP108, its upstream neurons (SMP353 and SMP354), or DANs using a recently developed dopamine indicator DA2m (Sun et al., 2020). With direct stimulation of DANs, release of dopamine was largely restricted to the compartment(s) innervated by Chrimson-expressing DANs (Figure 6-figure supplement1). Consistent with EM connectivity, activation of SMP108 or SMP353/354 evoked dopamine release in the

reward memory compartments $\beta'2$, $\gamma4$ and $\gamma5$ compartments (Figure 6). SMP108 activation also evoked small dopamine release in $\beta1$ and $\beta2$, presumably via indirect connections, but not in $\alpha1$. Notably, we observed that the dopamine signal in $\gamma2$, which is tuned to punitive stimuli, was significantly reduced after SMP108 activation (Figure 6figure supplement 1C). Other DANs for aversive memories such as PAM- $\gamma3$, PPL1- $\gamma1$ pedc, and PPL1- $\alpha3$ showed very weak response, if any. Thus, activation of SMP108 triggers dopamine release selectively in multiple reward memory compartments.

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311 SMP108 is required for second-order conditioning

312 As expected from above results, we found that blocking neurotransmission of SMP108 313 by expression of TNT using two different split-GAL4 drivers impaired second-order 314 conditioning compared to genetic controls (Figure 7A). We were unable to block 315 SMP108 only during the second-order conditioning using the thermogenetic effector shibire^{ts1} because flies with control genotype rapidly extinguished the first-order memory 316 317 and failed to perform second-order conditioning at the 32°C restrictive temperature (data 318 not shown). Nonetheless, blocking SMP108 with TNT did not impair the first-order 319 memory with 2min or 1-day retention (Figure 7B), indicating that flies with blocked 320 SMP108 were fully capable of smelling odors, tasting sugar, and forming, consolidating, 321 and retrieving the first-order appetitive memory.

To further assess the potential contribution of SMP108 to appetitive memory retrieval, we tested whether activation of SMP108 triggers any relevant behavior. Flies steer to an upwind orientation in the presence of reward-predicting odors and foodrelated odors like vinegar (Álvarez-Salvado et al., 2018; Borst and Heisenberg, 1982; Handler et al., 2019). Upon optogenetic stimulation of SMP108 with CsChrimson, flies indeed changed their mean orientation and walked upwind in the same circular arena used in the olfactory conditioning experiments described above (Figure 7-figure

329 supplement 1A). However, we did not observe any impairment of upwind steering in 330 response to the sugar-associated odor in SMP108-blocked flies (Figure 7-figure 331 supplement 1B), suggesting the existence of redundant circuits that trigger memory-332 based upwind steering. Thus, SMP108 could contribute to retrieval of reward memory for 333 guiding actions, but its requirement is limited to second-order conditioning. Taken 334 together, these results indicate that SMP108, which we identified as a prominent 335 anatomical hub for the feedforward circuit between reward memory compartments, 336 indeed plays a key role in second-order conditioning by triggering dopamine signals in 337 response to the reward-predicting cue.

338

339 SMP108 pathway induces transient memory

340 Based on the results so far, we propose a teacher-student compartment model that 341 explains the induction mechanism of second-order memory and its distinct dynamics 342 from first-order memory (Figure 8A). In this model, local plasticity induced in a stable 343 memory compartment (i.e. α 1) during first-order conditioning functions as a reinforcer to 344 induce secondary plasticity in other transient memory compartments through 345 interneurons (i.e. SMP108) that connect those memory compartments. Thus, this model 346 predicts that target compartments of SMP108 pathway collectively express transient 347 memory dynamics that recapitulates unstable nature of second-order memory induced 348 by sugar-odor (Figures 1F and 1G) or optogenetic conditioning (Figures 1H, 2E and 2G).

To test this prediction, we next examined the dynamics of memory induced by the SMP108 pathway in detail and compared them to those induced by direct stimulation of PAM- α 1 and other DAN types using CsChrimson (Figures 8B and Figure 8-figure supplement 1). The protocol started by assessing naïve odor preference that was designed to be canceled by reciprocal experiments. Then flies were sequentially trained five times by 10s, 30s, 60s, 60s and 60s periods of odor presentation paired with LED

355 activation, and then another odor presented without LED activation (training phase). 356 Memory was tested by giving a choice between odors after each training. After the fifth 357 training, memory was tested 12 times without pairing with LED activation (extinction 358 phase). Then flies were trained with a reversal protocol 5 times and tested 12 times 359 (reversal phase). After one more round of reversal phase (re-reversal), flies were 360 exposed to LED activation without odor to test the susceptibility of memory to non-361 contingent activation of DANs, a protocol that is known to erase memory (Berry et al., 362 2012; Placais et al., 2012). These experiments revealed that memories induced by 363 SMP108 or its upstream SMP353/354 differ in several ways from the memory induced 364 by activation of PAM- α 1 (Figure 8C-F). First, SMP108 and SMP353/354 can induce 365 memory more rapidly than PAM- α 1 (Figure 8C). Second, memories formed by SMP108 366 and SMP353/354 declined during later training sessions and during the extinction phase, 367 whereas memory formed by PAM- α 1 remained high (Figure 8D and E). Third, memory 368 formed by PAM- α 1 was resistant to DAN activation, but memories formed by SMP108 369 and SMP353/354 were decreased (Figure 8F). Such transient learning and fast 370 extinction are reminiscent of second-order conditioning by sugar (Figures 1F and 1G) or optogenetics (Figures 1H, 2E, and 2G). In contrast to the activation of CsChrimson in 371 372 PAM-α1, drivers that target CsChrimson to SMP108's downstream DANs exhibited 373 memory dynamics similar to those observed when CsChrimson is activated in SMP108 374 or SMP353/354. For instance, MB032B and MB213B split-GAL4 that target CsChrimson 375 in β' 2m and $\beta 1/\beta 2$, respectively, induced transient memories (Figure 8E). Consistent with 376 this, fitting the memory dynamics formed by SMP108 with a linear sum of direct DAN 377 activation data indicated an overweight of MB032B (β '2m), MB213B (β 1/ β 2) and 378 MB312C (γ 4), and zero weight for MB043C (α 1) (Figure 8G). However, the high memory 379 score of SMP108 activation after the first 10s training was fitted poorly, indicating that 380 combinatorial activation of DANs and/or suppression of DANs innervating y2 (Figure 6figure supplement 1C) might have a synergistic effect on memory formation. These experiments highlight the distinct memory properties exhibited by upstream and downstream partners of SMP108, and might help explain the circuit mechanisms underlying the difference between first- and second-order memories.

385

386 **Discussion**

In this study, we used the *Drosophila* mushroom body as a model system to examine how multiple dopamine-driven memory circuits interact to enable second-order conditioning. Although second-order conditioning has been demonstrated behaviorally in many species, there is little circuit-level knowledge to provide mechanistic insight. By developing a robust appetitive second-order conditioning protocol and utilizing the EM connectome map in *Drosophila*, we uncovered neural circuit mechanisms that define dynamics and learning rules of second-order conditioning.

394

Origins of the unique learning rules of second-order conditioning

396 Our optimization of the second-order conditioning protocol using actual sugar reward or its 397 optogenetic substitution revealed important properties of second-order memory and 398 enabled detailed circuit interrogation. Formation of second-order memory was most 399 effective either when the first-order S1 odor predicted a strong sugar reward (Figure 1D) or 400 when long-term first-order memory was optogenetically induced (Figure 2B). Furthermore, 401 during second-order training following optogenetic first-order conditioning, S2 odor must 402 precede the S1 odor (Figure 2C). With additional second-order training sessions, second-403 order memory could become as robust as the first-order memory, but the continual 404 omission of the expected fictive reward during training and extinction trials tended to 405 reduce second-order memory (Figures 1H, 2E, and 2G). The retention of second-order

406 memory was also shorter than first-order memory when we used actual sugar reward for 407 first-order conditioning (Figure 1F). Remarkably, all the dynamics and learning rules we 408 found in Drosophila for second-order conditioning are well-conserved across animal phyla 409 (Gewirtz and Davis, 2000; Pavlov, 1927; Rescorla, 1980). Our study indicates that, in flies, 410 at least some of these phenomena can be accounted for by the teacher-student model of 411 the MB circuit, which hypothesizes distinct dynamics of plasticity in individual 412 compartments and hierarchical interactions between compartments. Namely, a 413 compartment with a slow learning rate instructs compartment(s) with transient memory 414 dynamics.

415 Requirement of long first-order training for successful formation of second-416 order memory (Figure 1C and D) can be explained by the properties of the α 1, which we 417 identified as the teacher compartment. The DANs in $\alpha 1$ respond to sugar relatively weakly 418 compared to other DANs in the β'_2 , β_2 , γ_4 , γ_5 compartments (Siju et al., 2020). Also the 419 a1 compartment exhibited the slowest learning rate of all compartments even with 420 optogenetic stimulation of DANs that efficiently release dopamine (Figure 6-figure 421 supplement 1 and Figure 8C). Once established, however, memory in the α 1 is highly 422 resistant to extinction (Figures 2A and 8D), which is likely critical for forming second-order 423 conditioning without compromising first-order memory. These considerations emphasize 424 the eligibility of the α 1 compartment as a teaching compartment among all reward-memory 425 compartments. On the other hand, transient and unstable nature of second-order memory 426 can be ascribed to collective properties of student compartments (Figure 8). Future studies 427 are required to identify intrinsic molecular factors and microcircuit elements responsible for 428 distinct dynamics of teacher and student compartments.

429

430 Implications to the higher-order functions of heterogeneous dopamine subsystems

431 Our study identified a role of hierarchical interaction between dopamine-based memory 432 subsystems. Importantly, heterogeneous populations of DANs are also found in vertebrate 433 species, and they are involved in distinct types of learning. Studies using visual 434 conditioning in monkeys found that distinct types of DANs projecting to the head or tail 435 regions of the caudate nucleus change their response to reward-predicting cues with very 436 different dynamics (Kim et al., 2015, 2014). A recent study in rodents indicated that 437 subsets of DANs have diverse learning rates to compute positive and negative reward 438 prediction errors to enable distributional reinforcement learning (Dabney et al., 2020). Cue-439 evoked dopamine transients at the onset of reward-predicting cues are required for 440 second-order conditioning in rodents (Maes et al., 2020). Such dopamine transients could 441 be derived from memory encoded by the same DAN, other type(s) of DANs, or both, 442 depending on the architecture of feedback circuits. Given the conserved nature of second-443 order memory transiency across animal phyla, future studies in vertebrate models may 444 also reveal a hierarchical interaction between dopamine cell types with fast and slow 445 dynamics in second-order conditioning.

446 Second-order conditioning is merely one example of learning that depends on 447 higher-order connections between dopamine-dependent memory subsystems. In fact, in 448 flies, feedback and feedforward connections between MBONs and DANs or lateral 449 connections between MBONs are implicated in extinction of aversive and appetitive 450 memory as well as consolidation of memories (Felsenberg et al., 2018, 2017; McCurdy et 451 al., 2021). The EM connectome map, along with computational modeling (Gkanias et al., 452 2022; Jiang and Litwin-Kumar, n.d.), will guide further investigation of intercompartmental 453 interactions. For instance, we identified one outlier cell type of GABAergic interneuron 454 LHCENT3 that receives inputs from glutamatergic MBON- γ 5 β '2a and outputs to reward 455 DANs (Figure 4C). This cell type may serve as the substrate for subtraction of expected 456 reward in the computation of reward prediction error, as GABAergic neurons in VTA do in

457 vertebrate brains (Starkweather and Uchida, 2021). Although the majority of circuit-level research has focused on rather simple forms of learning that involve primary reinforcers, 458 459 animals have abundant opportunities to shape their behaviors through indirect learning 460 that depends on existing memory. We expect that network motifs similar to what we 461 identified here contribute to various forms of such complex learning. We expect that future 462 modeling studies constrained by the EM connectome and large scale behavioral and 463 neural activity data will lead to a comprehensive understanding of the MB's contributions 464 to these computations.

465

466 **Contents of second-order conditioning**

467 Understanding what is learned is a fundamental challenge in studies of associative 468 learning. There are many possible structures of associations that would allow animals to 469 perform second-order conditioning tasks. Our finding of the cross-compartmental nature of 470 second-order conditioning makes it unlikely that flies associate S2 with a specific type of 471 reward used as US, because individual MB compartments are tuned to different kinds of 472 rewards or reward responses. That is, while DANs in the teacher compartment $\alpha 1$ are 473 essential for nutritional value learning (Yamagata et al., 2015), those in the student 474 compartments y4 and β'^2 respond to water in thirsty flies (Lin et al., 2014). DANs in y4, y5 475 and β'^2 also represent vinegar and activity of DANs in γ^4 correlates with upwind steering 476 (Lewis et al., 2015; Zolin et al., 2021). DANs in β '2a also respond to a punishment-477 predicting odor when punishment is omitted (McCurdy et al., 2021). Thus, based on our 478 circuit mapping and the known functions of the relevant circuits, we propose that S2 is 479 associated with positive valence that was originally associated with S1 but generalized to 480 broader types of rewards. This view is consistent with the fact that second-order 481 conditioning is typically insensitive to subsequent reduction of the value of the US (i.e. 482 devaluation), which suggests that an association is formed between S2 and the original

483 valence of the US rather than the US itself (Rescorla, 1980). Studies in rodents 484 demonstrated that S1 and S2 with different sensory modalities can elicit distinct 485 conditioned responses (CRs), supporting the idea that S2 is not associated with the 486 specific CR elicited by S1 (Holland, 1977; Kim et al., 1996). Notably, a broadening of the 487 category of expected rewards in second-order conditioning has been suggested by a study 488 in pigeons (Stanhope, 1992), where differential CRs to qualitatively distinct USs (i.e. food 489 and water) were observed for S1 but not for S2. Thus, our circuit underpinning of second-490 order conditioning provides a concrete neuronal substrate for behavioral and psychological

491 phenomena that have been described for decades.

492 Materials and Methods

493 Fly strains

494 Drosophila melanogaster strains were reared at 22C and 60% humidity on standard 495 cornmeal food in 12:12 hour light:dark cycle. 4-10 days of adult females were used 2-4 496 days after sorting them on the Peltier cold plate. For flies expressing Chrimson (Klapoetke et al., 2014) the food was supplemented with retinal (0.2 mM all-trans-retinal prior to 497 498 eclosion and then 0.4 mM). Driver and effector lines are listed in the key resource table 499 and genotypes used by each figure are listed below. The new collection of split-GAL4 and 500 split-LexA drivers was designed based on confocal image databases 501 (http://flweb.janelia.org) (Jenett et al., 2012), and screening expression patterns of 502 p65ADZp and ZpGAL4DBD combinations as described previously (Aso et al., 2014; 503 Pfeiffer et al., 2010). Confocal stacks of new split-GAL4 driver lines used in this study are 504 available at http://www.janelia.org/split-gal4.

505

506 Detailed fly genotypes used by figures

Figure	Genotype
Figure 1C-G, Figure 1-figure supplement 1	Canton S
Figure 1H	w/w, 20xUAS-CsChrimson-mVenus attP18;+/Gr64f-GAL4;+/Gr64f-GAL4
Figure 2A-C	w/w, 20xUAS-CsChrimson-mVenus attP18;;+/MB043C-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/MB213B-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;;+/MB312C-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;MB109B/MB315C-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/ Empty-split-GAL4
Figure 2D	w/+;Empty-split-GAL4/UAS-TNT (II) w/+;MB196B/UAS-TNT (II) w/+;MB043C/UAS-TNT (II)
Figure 3 Figure 3-figure supplement 3	w/w,13XLexAop2-IVS-ChrimsonR-mVenus-p10 attP18, 20XUAS-syn21 mScarlet-opt-p10 su(Hw)attp8; SS01308-split- GAL4/MB043-split-LexA
Figure 3-figure supplement 1	 w/w, 13XLexAop2-IVS-ChrimsonR-mVenus-p10 attP18, 20XUAS-syn21 mScarlet-opt-p10 su(Hw)attp8; +/MB043-split-LexA w/w, 13XLexAop2-IVS-ChrimsonR-mVenus-p10 attP18, 20XUAS-syn21 mScarlet-opt-p10 su(Hw)attp8; MB319C-split-GAL4/MB043-split-LexA w/w, 13XLexAop2-IVS-ChrimsonR-mVenus-p10 attP18, 20XUAS-syn21 mScarlet-opt-p10 su(Hw)attp8; SS01308-split-GAL4/MB043-split-LexA w/w, 13XLexAop2-IVS-ChrimsonR-mVenus-p10 attP18, 20XUAS-syn21 mScarlet-opt-p10 su(Hw)attp8; SS01308-split-GAL4/MB043-split-LexA w/w, 13XLexAop2-IVS-ChrimsonR-mVenus-p10 attP18, 20XUAS-syn21 mScarlet-opt-p10 su(Hw)attp8; SS67221-split-GAL4/MB043-split-LexA
Figure 3-figure	w/w,13XLexAop2-IVS-ChrimsonR-mVenus-p10 attP18, 20XUAS-syn21 mScarlet-opt-p10 su(Hw)attp8; MB319C-split-

supplement 2	GAL4/MB043-split-LexA
Figure 4E	w/w, pJFRC200-10xUAS-IVS-myr::smGFP-HA in attP18; pJFRC225-5xUAS-IVS-myr::smGFP-FLAG in VK00005/SS67221- split-GAL4
Figure 4F	pBPhsFlp2::PEST in attP3;; pJFRC201-10XUAS-FRT>STOP>FRT-myr::smGFP-HA in VK0005, pJFRC240-10XUAS- FRT>STOP>FRT-myr::smGFP-V5-THS-10XUAS-FRT>STOP>FRT-myr::smGFP-FLAG in su(Hw)attP1/SS67221-split-GAL4
Figure 5	13XLexAop2 IVS p10 ChrimsonR mVenus trafficked in attP18/+; 58E02-LexAp65 in attP40/ VT026646-p65ADZp in attP40 (ss45234-split); pJFRC28-10XUAS-IVS-GFP-p10 in su(Hw)attP1 / VT029309-ZpGdbd in attP2 (ss45234-split)
Figure 6, Figure 6-figure supplement 1	 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MB043C-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MB213B-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MB109B-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MB109B-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MB109B-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MB315C-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MB312C-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MB312C-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MS312C-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/MS312C-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/S33917-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/S232917-split-GAL4 w/w, 10XUAS-Chrimson88-tdTomato attP18; 13F02-LexAp65 attP40; LexAop2-DA2m VK00005/S233917-split-GAL4
Figure 7	w/+;SS67221/+ w/+;SS67221/UAS-TNT (II) w/+;SS45234/+ w/+;SS45234/UAS-TNT (II) w/+;Empty-split-GAL4/TNT (II)SS67221/TNT
Figure 7-figure supplement 1A	w/w, 20xUAS-CsChrimson-mVenus attP18;+/ Empty-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/SS67221-split-GAL4
Figure 7-figure supplement 1B	w/+;SS67221/+ w/+; SS67221/UAS-TNT (II) w/+;Empty-split-GAL4/TNT (II)SS67221/TNT
Figure 8, Figure 8-figure supplement 1	w/w, 20xUAS-CsChrimson-mVenus attP18;+/+;+/MB043C-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/SS33917-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/SS67221-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/MB032B-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/MB315C-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+//HB315C-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/MB312C-split-GAL4 w/w, 20xUAS-CsChrimson-mVenus attP18;+/MB312C-split-GAL4

507 **Olfactory conditioning**

508 Olfactory conditioning was performed as previously described (Aso et al., 2016). Groups 509 of approximately 20 females of 4-10 d post-eclosion were trained and tested using the 510 modified four-field olfactory arena (Aso and Rubin, 2016; Pettersson, 1970) equipped 511 with the 627nm LED board (34.9 μ W/mm2 at the position of the flies) and odor mixers. 512 The flow rate of input air from each of the four arms was maintained at 100 mL/min 513 throughout the experiments by mass-flow controllers, and air was pulled from the central 514 hole at 400 mL/min. Odors were delivered to the arena by switching the direction of 515 airflow to the tubes containing diluted odors using solenoid valves. The odors were 516 diluted in paraffin oil: 3-octanol (OCT 1:1000), 4-methylcyclohexanol (MCH; 1:750), 517 Pentyl acetate (PA: 1:10000) and ethyl lactate (EL: 1:10000). Sugar conditioning was performed by using tubes with sucrose absorbed Whatman 3 MM paper as previously 518 described (Krashes and Waddell, 2008; Liu et al., 2012). Before conditioning, flies were 519 520 starved for 40-48 hour on 1% agar. Videography was performed at 30 frames per second and analyzed using Fiji. For experiments with one day retention, flies were kept 521 522 in agar vials at 21C after first-order conditioning. For testing olfactory memories, 523 distribution of flies in four quadrants were measured for 60 s. The performance index 524 (PI) is defined as a mean of [(number of flies in the two diagonal quadrants filled the one 525 odor) - (number of flies in other two quadrants filled with another odor or air)]/(total 526 number of flies) during final 30 s of 60 s test period. The average PI of reciprocal 527 experiments is shown in figures to cancel out potential position bias and innate odor 528 preference. Although genotypes of flies were not hidden to experimentalists, handling 529 was minimized by automation of stimulus delivery. We included all the data if 530 experiments were validated by metadata such as airflow readout from the mass flow 531 controllers.

532 **Optimization of second-order conditioning**

533 To establish a training protocol for robust olfactory second-order conditioning in 534 Drosophila, we first characterized how innate preference for an odor (when compared 535 with pure air) changes over multiple trials using the four-armed olfactory arena (Figurefigure supplement 1)(Aso and Rubin, 2016; Pettersson, 1970). We previously chose 536 537 concentrations of two conventional odors, 4-methylcyclohexanol (MCH) and 3-octanol 538 (OCT), so that naïve fed flies show behavioral responses to each odor at a similar level, 539 minimizing bias between them (Tully and Quinn, 1985). At the same concentration, 540 starved flies showed slight attraction to the MCH at the first trial, then gradually shifted to 541 aversion in subsequent trials (Figure 1-figure supplement 1). In contrast, both fed and 542 starved flies showed aversion to the OCT, which gradually decreased in subsequent 543 trials. Because the innate aversiveness of OCT may preclude appetitive second-order 544 conditioning, we decided to use MCH as the first conditioned stimulus (S1) throughout 545 this study.

546 The strength of second-order conditioning tends to be low, compared to that of 547 first-order, but can be enhanced by using an unconditioned stimulus (US) of high 548 intensity and sensory stimuli within the same modality (Helmstetter and Fanselow, 1989; 549 Rescorla and Furrow, 1977). Thus, we examined the effect of increasing conditioning 550 duration. After pairing MCH with sugar for increasing durations (0, 2, 5 min), flies were 551 allowed to consolidate the memory for one day. Then the stability of first-order memory was tested by repeating binary choice between S1 odor and air for 12 times. All trained 552 553 flies showed attraction to MCH during at least the first five trials (Figure 1C). One 2-min 554 training was enough to induce appetitive memory (Krashes and Waddell, 2008; Tempel 555 et al., 1983), but longer 5-min training resulted in slightly stronger memories during the first five tests on average. Therefore, we decided to limit the number of second order 556 557 conditioning to five times. We used two odorants, pentyl acetate (PA) and ethyl lactate 558 (EL) as the second conditioned stimuli (S2). These odors are known to evoke discrete 559 patterns of activity in Kenyon cells (Campbell et al., 2013) and thought to be easily 560 discriminated against. Innate behavioral responses to these odors were relatively stable 561 over 12 trials (Figure 1-figure supplement 1).

562 For first-order conditioning, flies learn best when sensory cues precede US or 563 DAN activation (Aso and Rubin, 2016; Tanimoto et al., 2004). Thus, during second-order 564 conditioning, 20 seconds of one S2 odor (S2+) was immediately followed by 10 seconds 565 of the S1 odor, whereas another S2 odor (S2-) was presented alone. Flies failed to form 566 second-order memory when S1 preceded S2+ (Figure 2C). PA and EL were S2+ and 567 S2- odors, respectively, in half of a set of reciprocal experiments. The S2+ and S2-568 odors were swapped in the other half of reciprocal experiments. After five training sessions, unpaired control flies showed weak attraction to S2+, possibly due to innate 569 570 attractiveness of MCH in starved flies (Figure 1-figure supplement 1). Compared to this 571 basal response, flies preferred the S2+ odor over the S2- odor when first-order 572 conditioning was long enough (i.e. 5min; Figure 1D). This preference for the S2+ odor 573 was not due to stimulus generalization of S1 (MCH) to PA or EL, because such bias is 574 designed to be canceled by our experimental design involving reciprocal experiments. 575 Both immediate and one-day first-order memories were potent to induce second-order 576 memory, but second-order memory did not last for one day (Figure 1F.

577 **Response Airflow**

For testing airflow directional response, we used the same circular olfactory arena (Figure 7-figure supplement 1), in which air flows from peripheral to a hole at the center. Each fly's distance from center (r_i) was measured and area normalized index (r_i/r_{arena})*(r_i/r_{arena}) was calculated. r_{arena} is the radius of the arena. When flies distribute randomly in the arena, mean r is 1/sqrt(2) and area normalized index is 1/2. To calculate upwind displacement, the mean of arena normalized distance from center at each time point in each movie was subtracted by that at the onset of LED or odor.

585

586 Electrophysiology

587 Fly stocks for electrophysiological experiments were maintained at room temperature on 588 conventional cornmeal-based medium (Archon Scientific). Experimental flies were 589 collected on the day of eclosion, transferred to all-trans-retinal food (0.5 mM) and kept in 590 the dark for 48-72 hr. For second-order conditioning experiments, flies were starved for 591 60-72 hr after feeding retinal food.

592 In vivo whole-cell recordings were performed as previously reported (Hige et 593 al., 2015). The patch pipettes were pulled for a resistance of 4-6M Ω and filled with pipette 594 solution containing (in mM): L-potassium aspartate, 140; HEPES, 10; EGTA, 1.1; CaCl₂, 595 0.1; Mg-ATP, 4; Na-GTP, 0.5 with pH adjusted to 7.3 with KOH (265 mOsm). The 596 preparation was continuously perfused with saline containing (in mM): NaCl, 103; KCl, 3; 597 CaCl₂, 1.5; MgCl₂, 4; NaHCO₃, 26; N-tris(hydroxymethyl) methyl-2-aminoethane-sulfonic 598 acid, 5; NaH₂PO₄, 1; trehalose, 10; glucose, 10 (pH 7.3 when bubbled with 95% O₂ and 599 5% CO₂, 275 mOsm). For recordings from starved flies, trehalose and glucose were 600 replaced by equimolar sucrose. Whole-cell recordings were made using the Axon 601 MultiClamp 700B amplifier (Molecular Devices). Target cells were visually targeted by 602 fluorescence signal with a 60X water-immersion objective (LUMPlanFl/IR; Olympus) 603 attached to an upright microscope (OpenStand; Prior Scientific). Cells were held at around 604 -60 mV by injecting hyperpolarizing current, which was typically < 100 pA. Signals were 605 low-pass filtered at 5 kHz and digitized at 10 kHz.

For odor delivery, a previously described custom-designed device was used (Hige et al., 2015). Saturated head space vapors of pure chemicals were air-diluted to 0.5 % (for second-order conditioning) or 2% (for the other experiments) before being presented to flies. Photostimulation was delivered by a high-power LED source (LED4D067; Thorlabs) equipped with 625 nm LED. Light pulses controlled by an LED driver (DC4100; Thorlabs) were presented to the brain at 17 mW/mm² through the objective lens.

Data acquisition and analyses were done by custom scripts in MATLAB (MathWorks). Instantaneous spike rates were calculated by convolving spikes with a Gaussian kernel (SD = 50 ms). Subthreshold odor responses and odor-evoked spikes were calculated with the time window of 1.2 s (for 1-s odor presentation) or 20.6 s (for 20-s odor presentation) from odor onset. Spontaneous spikes were subtracted to calculate odor-evoked spikes.

619

620 **Dopamine imaging**

Virgin females of *10XUAS-Chrimson88-tdTomato attP18; R13F02-LexAp65 in attP40;LexAop2-DA2m in VK00005* (Klapoetke et al., 2014; Sun et al., 2020) were crossed with split-GAL4 driver lines, and progenies were reared at 25 °C on retinal supplemented (0.2 mM) cornmeal medium that was shielded from light. All experiments were performed on female flies, 3-7 days after eclosion. Brains were dissected in a saline bath (103 mM NaCl, 3 mM KCl, 2 mM CaCl2, 4 mM MgCl2, 26 mM NaHCO3, 1 mM NaH2PO4, 8 mM trehalose, 10 mM glucose, 5 mM TES, bubbled with 95% O2 / 5% CO2). After dissection, the brain was positioned anterior side up on a coverslip in a Sylgard dish submerged in 3 ml saline at 20°C. The sample was imaged with a resonant scanning 2-photon microscope with near-infrared excitation (920 nm, Spectra-Physics, INSIGHT DS DUAL) and a 25× objective (Nikon MRD77225 25XW). The microscope was controlled using ScanImage 2016 (Vidrio Technologies). Images were acquired over a 231 μ m × 231 μ m x 42 μ m volume with a step size at 2 μ m. The field of view included 512 × 512 pixel resolution taken at approximately 1.07 Hz frame rate. The excitation power during imaging was 19 mW.

635

636 For the photostimulation, the light-gated ion channel CsChrimson was 637 activated with a 660-nm LED (M660L3 Thorlabs) coupled to a digital micromirror device 638 (Texas Instruments DLPC300 Light Crafter) and combined with the imaging path with a 639 FF757-DiO1 dichroic (Semrock). On the emission side, the primary dichroic was Di02-R635 640 (Semrock), the detection arm dichroic was 565DCXR (Chroma), and the emission filters 641 were FF03-525/50 and FF01-625/90 (Semrock). An imaging session started with a 30 s 642 baseline period, followed by a 1 s stimulation period when 12 µW/mm2 photostimulation 643 light was delivered, and responses were detected over a 30 s post stimulation period. This 644 was repeated for 10 trials. The light intensity was measured using the Thorlabs S170C 645 power sensor.

646 For quantification of dopamine sensor signals, we used custom python scripts 647 to draw ROIs corresponding to mushroom body compartments on maximum intensity 648 projection over time. Before calculating the change in fluorescence (ΔF), fluorescence from 649 a background ROI was subtracted. The background ROI was drawn in a region with no 650 fluorescence. Baseline fluorescence is the mean fluorescence over a 30 s time period 651 before stimulation started. The ΔF was then divided by baseline to normalize signal ($\Delta F/F$). 652 The mean responses from the 10 trials were calculated for each animal (4-6 samples per 653 driver). Kruskal-Wallis H (KW) test was used for multi-comparison. Post-hoc pairwise 654 comparison was made with the Wilcoxon rank-sum test.

655

656 **Connectivity analysis**

657 For producing the connectivity data shown in Figures 4 and Figure 4-figure supplement 1-3, 658 connectivity information was retrieved from neuPrint (neuprint.janelia.org) hosting the 659 "hemibrain" dataset (Scheffer et al., 2020), which is a publicly accessible web site 660 (https://doi.org/10.25378/janelia.12818645.v1). For cell types, we cited cell type assignments reported in Sheffer et al., 2020. Only connections of the cells in the right 661 hemisphere were used due to incomplete connectivity in the left hemisphere (Zheng et al., 662 663 2018). Connectivity data was then imported to а software Cytoscape 664 (https://cytoscape.org/) for generating the diagrams before finalizing on Illustrator. The 3D 665 renderings of neurons presented were generated using the visualization tools of NeuTu (Zhao et al., 2018) or VVD viewer (https://github.com/takashi310/VVD Viewer; (Wan et al., 666 667 2012).

668

669 Neurotransmitter prediction

670 The method for neurotransmitter prediction using electron microscopy images and a 3D 671 VGG-style network were described in detail for the FAFB data of a whole fly brain (Eckstein 672 et al., 2020; Zheng et al., 2018). We used the same approach to train the network to classify individual presynaptic sites of FIB-SEM hemibrain data into the same six major 673 674 neurotransmitters in fly brains as for FAB, i.e.: GABA, glutamate, acetylcholine, serotonin, 675 dopamine and octopamine. Due to the differences in resolution between FAFB and the 676 electron microscopy images used here, we adapted the architecture of the 3D VGG 677 network to be isotropic as follows: We use four downsampling layers with uniform pooling

678 sizes of 2x2x2 on 3D crops centered on synapses with a side-length of 80 voxels. The 679 results for 396 major interneurons are summarized in Supplementary File 1.

680

681 Immunohistochemistry

Brains and ventral nerve cord of 4-10 days old female were dissected, fixed and immunolabeled as previously described using the antibodies listed in the Key Resource Table (Aso et al., 2014; Nern et al., 2015). Samples were imaged with confocal microscopes (Zeiss LSM710, LSM780 or LSM880). Inset images in Figure 4E were taken with Airyscan.

687

688 Regression analysis of SMP108 memory dynamics

For each strain, the log-probability ratio of reinforced vs. unreinforced stimuli was computed as R = log(p/(1-p)), where *p* is the probability of choosing the reinforced stimulus. To relate the memory dynamics induced by SMP108 to those induced by DANs that it activates, we performed non-negative linear least-squares regression of the log-probability ratio for SMP108 against the ratios for PAM DANs. This reflects an assumption that the combinatorial activation of multiple compartments contributes a behavioral bias that is additive in log-probability ratio.

696

697 Statistics

698 Statistical comparisons were performed on GraphPad Prism or MATLAB using the 699 Kruskal Wallis test followed by Dunn's post-test for multiple comparison, t-tests, or two-700 way ANOVA followed by Tukey's post hoc multiple comparisons test designated in figure 701 legends. Non-parametric test was preselected for behavioral assays due to expected 702 lack of normality or equal variance in subsets of data. Sample size was not 703 predetermined based pilot experiments.

704 **Data availability**

- The confocal images of expression patterns are available online
- 706 (http://www.janelia.org/split-gal4). The source data for each figure are included in the
- 707 manuscript.

708 Supplemental information

709 Supplementary File 1 Neurotransmitter prediction and a full connection matrix for 710 MBONs, DANs and 396 interneurons cell types.

Numbers in column B-G are numbers of presynaptic sites that are predicted to be designated neurotransmitters. EM id in column K is an identification number in EM hembrain data. The other columns are the connection matrix. Top row indicates the direction of connections. For instance, 153 in the raw 5 of column M indicate the number of connections from MBON01 to SMP108, while 166 in the raw5 of column BD indicate the number of connections from SMP108 to PAM02. For the cell type consisting of multiple cells, a summed number of connections are shown.

719 Appendix

720 Key Resource Table

- 721
- 722

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737

738 Competing interests

- The authors declare no competing interests.
- 740
- 741

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1023 Figure legends

1024 Figure 1. Appetitive olfactory second-order conditioning in Drosophila

1025 (A) A simplified diagram of the mushroom body circuit. Identity of odors are encoded by 1026 patterns of activity in ~2,000 Kenyon cells. Contingent activity of Kenyon cells and 1027 dopamine release leads to plasticity of excitatory synapses from Kenyon cells to MB 1028 output neurons with compartment-specific dynamics.

(B) A diagram of the four-armed olfactory arena. Flies were confined in the 9 cm diameter circular area above the LED board. For odor-sugar conditioning, flies were first trained in a tube by pairing an odor with dried sugar paper, and then introduced to the olfactory arena. Performance index was calculated by counting the number of flies in each quadrant (see Methods).

1034 (C) Dynamics of MCH preference after various 2 or 5 min of first-order conditioning with
1035 sugar. Flies were trained after 40-48 hours of starvation and memories were tested 201036 24 hours later without feeding in between by examining preference to MCH over air for
1037 12 times. Unpaired group received 5 min of sugar 2 min prior to 5 min exposure to MCH.
1038 Mean performance index of the first 5 tests after 5 min training was higher than that of 2

- 1039 min. p<0.01; unpaired t-test; N=10-12.
- (D) Second-order memory performance by wild type flies. n.s., not significant (p=0.152);
 ***, p<0.0001; Dunn's multiple comparison tests following Kruskal-Wallis test; N = 14–16.
 Means and SEMs are displayed with individual data points.

1043 (E) The odor preference following the sensory preconditioning protocol, in which the 1044 order of the first and second-order conditioning was swapped. n.s., not significantly 1045 different from the chance level; Wilcocxon signed-rank test; N=12.

1046 (F) Retention of second-order memory. After 24-hour, the second-order memory 1047 decayed to the chance level. ***, p<0.001; Wilcocxon signed-rank test or Mann-Whiteney 1048 test; N = 12.

(G) Odor preference between two S2 odors after the second-order or first-order conditioning was measured for six times by alternative position of two odorants with 2 min intervals. Memory persistency, a mean of PIs for 3rd-6th tests divided by PI of 1st test, was significantly smaller for second-order memory. **; p<0.0022; Mann-Whitney test; N=6. Means and SEMs are displayed with individual data points.

1054 (H) Learning curves by first-order, second-order, or second-order without omission of 1055 optogenetic reward. Flies expressing CsChrimson in sugar sensory neurons with Gr64f-1056 GAL4 were trained by pairing S2+ odor with activation of LED (First) or S1 odor that was 1057 previously paired with LED (Second). In the no omission protocol, sugar sensory 1058 neurons were activated immediately after S1 by repeating 1s red LED illumination with 1059 1s intervals for three times. Preference between S2+ and S2- odors was tested after 1st, 3rd. 5th. 7th and 9th training sessions. After 9th training, memory by second-order 1060 1061 protocol was lower than other protocols and its peak at 3rd training (p<0.05); Dunn's 1062 multiple comparison tests following Kruskal-Wallis test; N=8.

- (I) Learning of S2 odors was compromised when S1 odor paired with Gr64f>CsChrimson
 precedes S2+ odor. *, p<0.05 by Dunn's tests following Kruskal-Wallis test; N=12.
- 1065
- 1066 Figure 1-soure data 1
- 1067 1068

Figure 1-figure supplement 1 Dynamics of odor preference

- 1070 (A)Twelve repetition of odor preference of fed and 40-48 hour starved flies.
- PIs represent results of reciprocal experiments. The half groups of flies went through the identical tests but with alternating positions of odors and air quadrants to cancel out potential positional preference for each rigs. The mean odor preferences during 12 tests were significantly different between fed and starved flies for MCH and EL. p<0.01; unpaired t-test; N=8. Means and SEM are shown.
- 1076 (B) Delta between the first test and subsequent tests in fed (left) and 40-48 hour starved
- 1077 flies (right). Areas under curve for OCT was significantly higher than that for other odors
- 1078 in both fed and starve flies, whereases the area under the curve for MCH was lower than
- 1079 other odors only in starved flies. p<0.01; unpaired t-test; N=8.
- 1080 Figure 1-figur supplement 1-soure data 1

1081 Figure 2. Identification of the teacher compartment(s)

- (A) Dynamics of S1 odor (MCH) preference after pairing 1 min of S1 odor with activation
 of different PAM cluster DANs for three times. Numbers of CsChrimson-mVenus in each
 driver per hemisphere and total number of corresponding DAN cell types in EM
 hemibrain data are indicated. At 3rd-7th tests, MCH preference of MB043C>CsChrimson
 flies was higher than all other genotypes. p<0.05; Dunn's multiple comparison tests
 following Kruskal-Wallis test; N=6.
- 1088 (B) The second-order conditioning 2-min or 1 day after the first-order conditioning with 1089 optogenetic activation of various DAN types. Second-order memory was tested 1090 immediately after pairing S2+ odor with S1 odor (MCH) five times. n.s., not significant; *, 1091 p=0.0330; **, p=0.0046 ***, p<0.001 ; Dunn's multiple comparison tests following 1092 Kruskal-Wallis test; N=8-10.
- 1093 (C) The second-order memory immediately after backward second-order conditioning.
 1094 Flies expressing CsChrimon-mVenus by MB043C split-GAL4 were trained with identical
 1095 protocol as in B, except that the onset of S1 odor was shifted to the 10 second before
 1096 the onset of the first S2 odor. n.s., not significant from zero; Wilcocxon signed-rank test;
 1097 N=6.
- 1098 (D) Preference to the S1 odor (left) and second-order memory (right) by flies expressing 1099 TNT with empty, MB196B or MB043C split-GAL4. MB196B labels ~27 cells per 1100 hemisphere, including PAM- γ 4, PAM- γ 4< γ 1 γ 2, γ 5 and β '2a. *, p=0.0126; ***, p<0.001; 1101 Dunn's multiple comparison tests following Kruskal-Wallis test; N=8 for S1 preference;
- 1102 N=10-14 for second-order.

1103 (F) Learning curves by first-order, second-order, or second-order without omission of

- 1104 optogenetic reward. Flies expressing CsChrimson with MB043C split-GAL4 were trained
- 1105 by pairing S2+ odor directly with optogenetic activation of DANs (First) or S1 odor that

1106 was previously paired with DAN activation (Second). In the no omission protocol, DANs 1107 were activated immediately after S1 by repeating 1s red LED illumination with 1s

1108

intervals for three times. Preference between S2+ and S2- odors was tested after 1st, 1109

3rd, 5th, 7th and 9th training sessions. After 9th training, memory by second-order 1110 protocol was lower than other protocols and its peak at 5th training. **, p<0.01; Dunn's

1111 multiple comparison tests following Kruskal-Wallis test; N=8-10.

1112 (G) The preference for the S1 odor (MCH) after the 9th session of second-order

- 1113 conditioning as in F. n.s., not significant; Mann-Whitney test; N=8.
- 1114 (H) Comparison of memory decay after repetitive tests. Flies were trained five times with
- 1115 first or second-order conditioning protocol as in F but without tests. Immediately after the
- 1116 5th training, preference between two S2 odors was measured repeatedly without training.
- 1117 At third test, second-order memory was significantly lower than first-order memory. **.
- p=0.0036; Dunn's multiple comparison tests following Kruskal-Wallis test; N=8. 1118
- 1119

1120 Figure 2-source data 1

1121

1122 Figure 3. Second-order Conditioning Induces Cross-compartmental Plasticity.

1123 (A) Experimental design and protocol. ChrimsonR-mVenus was selectively expressed in 1124 PAM-α1 using MB043-split-LexA (58E02-ZpLexADBD in JK22C; 32D11-p65ADZp in 1125 JK73A; see Figure 3-figure supplement 1 for expression pattern), and in vivo whole-cell 1126 recordings were made from MBON- γ 5 β '2a, which was labeled by mScarlet using a split-1127 GAL4 driver SS01308. For the first-order conditioning, 1-min presentation of S1 (MCH) 1128 was paired with LED stimulation (1 ms, 2 Hz, 120 times), which caused odor-specific 1129 suppression of responses in MBON- α 1 (Figure 3-figure supplement 2). After repeating 1130 first-order conditioning three times with 2-min intervals, second-order conditioning was 1131 performed by presenting S2+ (either PA or EL) for 20 s, and then S1 for 10 s with 5-s 1132 delay. S2- was presented alone 2 min later. Second-order conditioning was repeated 1133 five times, and the responses to S2 were recorded. In control experiments, first-order 1134 conditioning was performed in the same manner, but the presentation of S1 was omitted 1135 during second-order conditioning. Reciprocal experiments were performed by swapping 1136 S2+ and S2- in separate flies.

1137 (B) Mean responses (± SEM in light colors) to S2+ and S2- in the first (black) and fifth 1138 trials (red) during second-order conditioning (n = 14, including reciprocal experiments). 1139 Horizontal gray bars indicate 20-s odor presentation period.

1140 (C) Mean response magnitudes (± SEM) evoked by S2+ and S2-. The response

- 1141 magnitude was calculated by averaging the depolarization during the response window 1142 (0–20.6 s from odor onset). Each solid (PA used as S2+; n = 7) and dashed line (EL as 1143 S2+; n = 7) indicates data from a single fly. Responses to S2+ underwent depression 1144 after the first trial, while those to S2- did not change. Different letters indicate significant 1145 differences detected by Tukey's post hoc multiple comparisons test (p < 0.05) following 1146 repeated-measures two-way ANOVA (p = 0.003). There was no significant change in the
- 1147 peak amplitude (p = 0.87).

1148 (D, E) Same as (B) and (C) except that the data are from control experiments (n = 4)1149 each with PA or EL used as S2+, respectively). Neither responses to S2+ nor S2-1150 changed (p = 0.28; repeated-measures two-way ANOVA). The peak response did not 1151 change either (p = 0.22).

- 1152
- 1153

1154Figure 3-figure supplement 1 Expression patterns of MB043-split-LexA, MB319C1155and SS67221-split-GAL4

1156 (A) MB043-split-LexA drove expression ChrimsonR-mVenus in 4.7 PAM- α 1 neurons on average: (4, 4, 5, 4, 5, 6) cells per hemisphere were observed in three brain samples.

(B-C) The expression pattern of MB043-split-LexA was unaffected in the presence of MB319C-split-GAL4. We observed a few additional mScarlet-positive cells in addition to the two MBON-α1, presumably because interference between the AD hemi-driver of the MB043-split-LexA and DBD hemi-driver of MB319C-split-GAL4. For electrophysiology, MBON-α1 was found based on their soma location, brightness of mScarelt signals and odor response.

- 1164 (D-E) Expression patterns of MB043-split-LexA and SS67221-GAL4-split-GAL4. 1165 Additional cells expressed mScarlet in optic lobes, but MBON-γ5β'2a was 1166 unambiguously labeled by mScarlet in the central brain.
- 1167 (F-G) Expression patterns of MB043-split-LexA and SS67221-GAL4-split-GAL4 did not interfere.
- 1169

1170Figure 3-figure supplement 2 Optogenetic Conditioning in α1 Compartment1171Induces Depression in MBON-α1

- 1172 (A) Experimental design and protocol. ChrimsonR-mVenus was selectively expressed in 1173 PAM- α 1 using MB043-split-LexA, and in vivo whole-cell recordings were made from 1174 MBON- α 1, which was labeled by mScarlet using a split-GAL4 driver MB319C. 1-min 1175 presentation of CS+ (OCT or MCH) was paired with LED stimulation (1 ms, 2 Hz, 120 1176 times), followed by 1-min presentation of CS- alone. Reciprocal experiments were 1177 performed by swapping CS+ and CS- in a separate set of flies.
- (B) Membrane voltage (upper panels) and spike data (lower panels) from a single representative fly, in which OCT was used as CS+. Gray bars indicate 1-s odor presentation.
- 1181 (C) Time courses of instantaneous spike rate (mean \pm SEM; n = 6 and 5 for each set of experiment).
- 1183 (D) Summary data of mean odor-evoked spike counts (\pm SEM). Gray lines indicate data 1184 from individual neurons. After each pairing, responses to CS+ were suppressed, while 1185 those to CS- were either showed less suppression than CS+ or no change (repeated-1186 measures two-way ANOVA followed by Tukey's post hoc multiple comparisons test; *p < 1187 0.05, **p < 0.005, ***p < 0.001).
- 1188

11891190 Figure 3-figure supplement 3. Response to current injection in MBON-γ5β'2a

1191 Representative somatic voltage responses to current injection in MBON- $\gamma 5\beta'2a$ with 1192 (right) or without (left) TTX (1 mM). The current injection waveforms are shown in the 1193 inset in the middle. Depolarization typically increased the frequency of small, fast 1194 membrane potential fluctuations, which were partially suppressed by TTX. However, 1195 those events were not readily distinguishable from the putative synaptic potentials that 1196 were remaining in the presence of TTX. Therefore, it remains inconclusive whether 1197 MBON- $\gamma 5\beta'2a$ elicits action potentials.

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- 1200

1201 Figure 4. SMP108 is a key interneuron between MBON-a1 and DANs

(A)The connections from MBON-a1 to PAM cluster DANs with two interneurons
identified in the hemibrain EM data (Scheffer et al., 2020). The width of arrows indicate
number of connections. The colors of circles and arrows indicate type of putative
neurotransmitter. Single SMP353 and three SMP354s have similar morphology and

- projection patterns and converge on to SMP108. Cholinergic interneurons
 SMP353/SMP354 and SMP108 are shown as filled orange circles and arrows. Other
 cholinergic connections are shown in transparent orange. See Supplementary File 1 for
 a full connectivity matrix and neurotransmitter predictions. See Figure 4-figure
 supplement 2 for the SMP108's connections with subtypes of DANs.
- (B) Connections between the six neurons in the second layer in A and CRE011.
 SMP108 outputs to all three other putative cholinergic interneurons. LHPV10d1 is the top target of SMP108. SMP553 send its first and second strongest outputs to SMP108 and SMP177.
- 1215 (C) Total number of connections to reward DANs (PAM01, 02, 04, 06, 07, 08, 10, 11,15) 1216 which can induce appetitive memory with optogenetic activation, plotted against number 1217 of inputs from MBON- γ 5 β '2a. Each circle represents one of 396 interneuron cell types 1218 that have at least 100 total connections with MBONs and DANs. Similar to SMP108, 1219 CRE011 is an outlier cell type in terms of the high number of direct inputs from MBON-1220 γ 5 β '2a and outputs to reward DANs. See Figure 4-figure supplement 3 for other kinds of 1221 connections between these interneurons and DANs/MBONs.
- (D) A projection of a reconstructed SMP108 neuron in the hemibrain EM images aligned
 to a standard brain with outline of the brain and the MB lobes.
- 1224 (E) Confocal microscope images of SS67221 split-GAL4 driver with membrane-targeted 1225 reporter myr-smFLAG and presynaptic reporter Syt-smHA. Inset shows anti-ChAT 1226 immunoreactivity of SMP108's axon terminals. (F) Morphology of individual SMP108 1227 visualized by multi-color flip out of SS67221 split-GAL4.
- 1228

1229 Figure 4-figure supplement 1 Connections of MBON-α1 and SMP108

- 1230 (A) Numbers of direct MBON-to-DANs synaptic connections. White boxes indicate 1231 within-a-compartment connection such as connection from MBON-a1 to PAM-a1.
- 1232 (B) Total number of outputs to reward DANs (i.e. PAM01, 02, 04, 06, 07, 08, 10, 11, 15) 1233 which can induce appetitive memory upon optogenetic activation, plotted against total 1234 number of connection from MBON-a1 for 396 cell types that have at least 100 total 1235 connections with MBONs and DANs. Colors indicate predicted neurotransmitters. One of 1236 outstanding cell type, LHAD1b5, cannot mediate cross-compartmental pathways 1237 because it is exclusively connected with PAM-o1 but not other reward DANs. Another 1238 outstanding cell type LHAD1b2 d is a part of two-hop pathways from MBON-a1 to 1239 PAM-a1 (Figure 3A)
- 1240 (C) Total number of outputs to reward DANs plotted against total number of outputs to punishment DANs (i.e. PPL101, 103, 105, 106 and PAM12).
- 1242

1243 Figure 4-figure supplement 2 Connections from SMP108 to DAN subtypes

1244 The number in the top row is the total number of connections from the SMP108 to 1245 subtypes of DANs that were defined by their projection and connectivity patterns (Li et 1246 al., 2020). The number of cells per subtype and number of connections per cell are 1247 shown in the middle and bottom row, respectively.

1248

1249 Figure 4-figure supplement 3 Connections of interneurons with DANs and MBONs

Scatter plots display designated pairs of connections for 396 cell types that have at least
 100 total connections with MBONs and DANs in EM hemibrain data. Colors indicate
 predicted neurotransmitters. Names of outlier cell types are labeled. See Supplementary
 Table 1 for full data. Note that SMP354 and other cholinergic interneurons in 2-hop

pathway (Figure 4A), SMP177, LHPV5e1, LHAD1b2_d, LHPV10d1 showed a shared property of receiving converging inputs from glutamatergic MBONs and cholinergic MBONs, whose activity represent appetitive or aversive memories, respectively. In addition to input from MBON-α1, SMP354 receive converging inputs from MBON-α3, a compartment of long-term aversive memory.

1259 1260

1261 Figure 5. SMP108 acquires enhanced responses to reward-predicting odors

- (A) Experimental design and protocol. ChrimsonR-mVenus was expressed in PAM cluster DANs, which include PAM-α1, using R58E02-LexA. In vivo whole-cell recordings
 were made from SMP108, which was labeled by GFP using a split-GAL4 driver
- SS45234. In the first pairing (Pairing 1), 1-min presentation of OCT was paired with LED
 stimulation (1 ms, 2 Hz, 120 times), followed by 1-min presentation of MCH alone. Odors
 were flipped in the second round of pairing (Pairing 2). Responses to each odor (1-s
 presentation) were measured before (Pre) and after pairing 1 (Post 1), and after pairing
 2 (Post 2).
- 1270 (B) Membrane voltage (upper panels) and spike data (lower panels) from a single
- 1271 representative neuron. Gray bars indicate 1-s odor presentation.
- 1272 (C) Time courses of instantaneous spike rate (mean \pm SEM; n = 6).
- 1273 (D) Summary data of mean odor-evoked spike counts (± SEM). Gray lines indicate data
- 1274 from individual neurons. After each pairing, responses to paired odors were potentiated,
- 1275 while those to unpaired odors tended to decrease. Repeated-measures two-way ANOVA
- 1276 (p = 0.0001) followed by Tukey's post hoc multiple comparisons test. *p < 0.05.
- 1277

1278 Figure 6. SMP108 promotes dopamine release in multiple compartments

- (A) Representative images of Chrimson88-tdTtomato expression patterns (left) and maximum intensity projections of DA2m dF/F in the MB lobes (right). Release of dopamine upon activation of DANs or SMP108 pathways, measured with dopamine sensor DA2m expressed in Kenyon cells. *10XUAS-Syn21-Chrimson88-tdTtomato-3.1 in attP18* was driven with designated split-GAL4 driver lines. Fluorescence of DA2m in response to one second of 660nm LED light was measured in dissected brains with twophoton imaging of volume containing MB lobes (see Methods).
- (B) Mean DA2m dF/F in ROIs defined for each MB compartment. SEMs are shown asshading, although they are often within width of lines representing means. N=8-12.
- 1288 See Figure 6-figure supplement 1 for quantification and the data with direct simulation of 1289 DANs.
- 1290

1291Figure 6-figure supplement 1. Patterns of dopamine release by different driver1292lines

- 1293 Representative images of neurons expressing Chrimson88-tdTtomato by designated 1294 driver lines (left) and maximum intensity projection of DA2m dF/F in the MB lobes (right).
- 1295 (A) Representative images of Chrimson88-tdTtomato expression patterns (left) and max 1296 intensity projections of DA2m dF/F in the MB lobes (right) as in Figure 6A.
- 1297 (B) Mean DA2m dF/F in ROIs defined for each MB compartment. N=8-12.
- 1298 (C) Area under the curve during the 10s period after activation.
- 1299
- 1300 Figure 6-figure supplement 1 -source data 1
- 1301

1303 Figure 7. SMP108 is required for second-order memory

(A) Second-order memory immediately after 5 training sessions as in Figure 1D following
5min first order conditioning a day before. Blocking SMP108 by expressing TNT with
SS67221 or SS45234 impaired the second-order memory compared to genetic controls.
N=10-12.

(B) Preference to the S1 (MCH) odor over the air one day after pairing with sugar for5min. N=8-10.

1310 (C) First-order memory immediately after pairing S2+ odor with sugar for 2-min. N=8. *,

- p<0.05; **, p<0.01; Dunn's multiple comparison tests following Kruskal-Wallis test
- 1312
- 1313 Figure 7-source data 1
- 1314 1315

1316Figure 7-figure supplement 1 SMP108 can drive upwind steering but dispensable1317for the conditioned responses

1318 (A) Diagram of the circular arena (top). Airflow was constantly set at 400mL/min 1319 throughout the experiments, and 10s of 627nm LED stimulations was applied for six 1320 times with 2min intervals. Six trial averages of upwind displacement from the onset of 1321 LED (middle) and cosine of angle to upwind direction (bottom) are shown for flies (SMP108) 1322 SS67221 expressing CsChrimson or empty-split-GAL4. 1323 SS67221>CsChrimson flies showed enhanced upwind displacement (p<0.05) and 1324 orientation toward upwind during LED ON period (p<0.01); See the method for the 1325 calculation of upwind displacement. N=12.

(B) Groups of flies were trained by pairing either PA or EL with sugar for 2-min, and their
response to airflow in the presence of odors were examined 20-24 hours later. Flies
showed enhanced upwind displacement and orientation to upwind in the presence of
reward-predicting odor. Upwind steering of flies with blocked SMP108 (SS67221/UASTNT) was indistinguishable with control genotypes. N=15-16

1331

1332 Figure 7-figure supplement 1-source data 1

1333

1334Figure 8. SMP108 pathway induces transient memory

(A) Teacher-student compartments model of second-order conditioning hypothesizes
that "teacher" compartment with slow learning rate and persistent memory instructs other
compartments with faster learning rate and transient memory dynamics via
SMP353/SMP354 and SMP108.

(B) Dynamics of memory with optogenetic activation of SMP108 (SS67221),
SMP353/354 (SS33917) or various types of DANs. See texts and methods for
explanation of the protocol, and Figure 8-figure supplement 1 for specificity of
expression pattern in the central brain and the ventral nerve cord. Means and SEM are
displayed. N=8-14.

1344 (C) Learning rate defined as a (PI after first 10s training)/(peak PI during the first 51345 training trials) for each driver line.

(D) Persistency during training defined as (PI after 5th training)/(peak PI during the first5 training trials).

1348 (E) Persistency of memory defined as (mean of PIs during 12 tests after first training 1349 trials)/(peak PI during the first 5x training trials).

- 1350 (F) Resistance to DAN activation defined as (mean of last three tests following activation
- LED without odors)/(PI after 5th conditioning in re-reversal phase), which measures both transiency during training and extinction during 12 tests.
- p<0.05; **, p<0.01; ***, p<0.01; Dunn's multiple comparison tests following Kruskal-
 Wallis test; N=8-14.
- (G) The log-probability ratio of choosing the S2+ against S2- for SS67221 (SMP108)
 data were fitted best with weights of (0.57, 0.46,0.157,0,0,0) for data of DAN driver lines
 (MB032B, MB213B, MB312C, MB043C, MB109B and MB315C).
- 1358

1359 Figure 8-source data 1

1361 Figure 8-figure supplement 1. Expression patterns of drivers

- (A-H) Projection of confocal microscopy stacks for expression patterns of CsChrimson mVenus driven by designated split-GAL4 driver lines in brains and ventral nerve cords.
 Confocal stacks are available at https://splitgal4 ianelia.org
- 1364 Confocal stacks are available at <u>https://splitgal4.janelia.org</u>
- 1365 1366

Key Resources T	able			
Reagent type (species) or resource	Designation	Source or reference	Identifiers	Additional information
strain, strain background (Drosophila melanogaster)	Canton S	Martin Heisenberg	N.A.	
strain, strain background (Drosophila melanogaster)	20xUAS- CsChrimson -mVenus attP18	Klapoetke et al., 2014; PMID: 24509633	N.A.	
strain, strain background (Drosophila melanogaster)	10XUAS- Chrimson88- tdTomato attP1	Klapoetke et al., 2014; PMID: 24509633	N.A.	
strain, strain background (Drosophila melanogaster)	13XLexAop2 -IVS- ChrimsonR- mVenus-p10 attP18	Vivek Jayaraman	N.A.	
strain, strain background (Drosophila melanogaster)	20XUAS- syn21- mScarlet- opt-p10 su(Hw)attp8	Glenn Turner	N.A.	
strain, strain background (Drosophila melanogaster)	pJFRC200- 10xUAS- IVS- myr::smGFP -HA in attP18	Nern et al.,2015; PMID: 25964354	N.A.	
strain, strain background (Drosophila melanogaster)	pJFRC225- 5xUAS-IVS- myr::smGFP -FLAG in VK00005	Nern et al.,2015; PMID: 25964354	N.A.	

strain, strain background (Drosophila melanogaster)	pBPhsFlp2:: PEST in attP3	Nern et al.,2015; PMID: 25964354	N.A.	
strain, strain background (Drosophila melanogaster)	pJFRC201- 10XUAS- FRT>STOP >FRT- myr::smGFP -HA in VK0005	Nern et al.,2015; PMID: 25964354	N.A.	
strain, strain background (Drosophila melanogaster)	pJFRC240- 10XUAS- FRT>STOP >FRT- myr::smGFP -V5-THS- 10XUAS- FRT>STOP >FRT- myr::smGFP - FLAG_in_su (Hw)attP1	Nern et al.,2015; PMID: 25964354	N.A.	
strain, strain background (Drosophila melanogaster)	LexAop2- DA2m VK00005	Sun et al., 2020; PMID: 33087905	N.A.	
strain, strain background (Drosophila melanogaster)	MB043-split- LexA	This paper	N.A.	Available from Aso lab
strain, strain background (Drosophila melanogaster)	empty-split- GAL4 (p65ADZp attP40, ZpGAL4DB D attP2)	Seeds et al., 2014; PMID: 25139955	N.A.	
strain, strain background (Drosophila melanogaster)	MB032B split-GAL4	Aso et al., 2014a; PMID: 25535793	N.A.	

strain, strain background (Drosophila melanogaster)	MB043C split-GAL4	Aso et al., 2014a; PMID: 25535793	N.A.	
strain, strain background (Drosophila melanogaster)	MB109B split-GAL4	Aso et al., 2014a; PMID: 25535793	N.A.	
strain, strain background (Drosophila melanogaster)	MB213B split-GAL4	Aso et al., 2014a; PMID: 25535793	N.A.	
strain, strain background (Drosophila melanogaster)	MB315C split-GAL4	Aso et al., 2014a; PMID: 25535793	N.A.	
strain, strain background (Drosophila melanogaster)	SS33917 split-GAL4	This paper	N.A.	Available from Aso lab
strain, strain background (Drosophila melanogaster)	SS45234 split-GAL4	This paper	N.A.	Available from Aso lab
strain, strain background (Drosophila melanogaster)	SS67221 split-GAL4	This paper	N.A.	Available from Aso lab
strain, strain background (Drosophila melanogaster)	UAS-TeNT	Keller et al., 2002: PMID: 11810637	N.A.	
antibody	anti-GFP (rabbit polyclonal)	Invitrogen	A11122 RRID:AB_221 569	1:1000

antibody	anti-Brp (mouse monoclonal)	Developmental Studies Hybridoma Bank	nc82 RRID:AB_234 1866	1:30
antibody	anti-ChAT (mouse monoclonal)	Developmental Studies Hybridoma Bank	ChAT4B1 RRID:AB_528 122	1:50
antibody	anti-HA-Tag (mouse monoclonal)	Cell Signaling Technology	C29F4; #3724 RRID:AB_106 93385	1:300
antibody	anti-FLAG (rat monoclonal	Novus Biologicals	NBP1-06712 RRID:AB_162 5981	1:200
antibody	anti-V5-TAG Dylight-549 (mouse monoclonal)	Bio-Rad	MCA2894D549 GA RRID:AB_108 45946	1:500
antibody	anti-mous IgG(H&L) AlexaFluor- 568 (goat polyclonal)	Invitrogen	A11031 RRID:AB_144 696	1:400
antibody	anti-rabbit IgG(H&L) AlexaFluor- 488 (goat polyclonal)	Invitrogen	A11034 RRID:AB_257 6217	1:800
antibody	anti-mouse IgG(H&L) AlexaFluor- 488 conjugated (donkey polyclonal)	Jackson Immuno Research Labs	715-545-151 RRID:AB_234 1099	1:400
antibody	anti-rabbit IgG(H&L) AlexaFluor- 594 (donkey polyclonal)	Jackson Immuno Research Labs	711-585-152 RRID:AB_234 0621	1:500

antibody	anti-rat IgG(H&L) AlexaFluor- 647 (donkey polyclonal)	Jackson Immuno Research Labs	712-605-153 RRID:AB_234 0694	1:300
antibody	anti-Mouse IgG (H&L) ATTO 647N (goat polyclonal)	ROCKLAND	610-156-121 RRID:AB_108 94200	1:100
antibody	anti-rabbit IgG (H+L) Alexa Fluor 568 (goat polyclonal)	Invitrogen	A-11036 RRID:AB_105 63566	1:1000
chemical compound, drug	3-Octanol	Sigma-Aldrich	218405	
chemical compound, drug	4- Methylcyclo hexanol	VWR	AAA16734- AD	
chemical compound, drug	Pentyl acetate	Sigma-Aldrich	109584	
chemical compound, drug	Ethyl lactate	Sigma-Aldrich	W244015	
chemical compound, drug	Paraffin oil	Sigma-Aldrich	18512	
software, algorithm	lmageJ and Fiji	NIH Schneider et al., 2012	https://imagej.ni h.gov/ij/ http://fiji.sc/	
software, algorithm	MATLAB	MathWorks	https://www.m athworks.com /	

software, algorithm	Adobe Illustrator CC	Adobe Systems	https://www.a dobe.com/pro ducts/illustrat or.html	
software, algorithm	GraphPad Prism 9	GraphPad Software	https://www.gr aphpad.com/s cientific- software/pris m/	
software, algorithm	Python	Python Software Foundation	https://www.p ython.org/	
software, algorithm	neuPrint	HHMI Janelia	<u>https://doi.org/</u> <u>10.25378/jane</u> <u>lia.12818645.</u> <u>v1</u>	
software, algorithm	Cytoscape	(Shannon et al., 2003)	https://cytosc ape.org/	
software, algorithm	NeuTu	<u>Zhao et al.,</u> <u>2018</u>	https://github. com/janelia- flyem/NeuTu	
software, algorithm	ScanImage	Vidrio Technologies	https://vidriote chnologies.co m/	
software, algorithm	VVDveiwer	HHMI Janelia	<u>https://github.</u> <u>com/takashi3</u> <u>10/VVD_View</u> <u>er</u>	
other	Grade 3MM Chr Blotting Paper	Whatmann	3030-335	Used in glass vials with paraffin- oil diluted odours
other	mass flow controller	Alicat	MCW- 200SCCM-D	Mass flow controller used for the olfactory arena











Figure 2



Figure 3







Figure 3-figure supplement 2



Figure 3-figure supplement 3



Figure 4



Figure 4-figure supplement 1

DAN cell type	PAM01_a	PAM01_b	PAM15_a	PAM15_b	PAM02	PAM03	PAM04_a	PAM04_b	PAM05	PAM06_a	PAM06_b	PAM07	PAM08_a	PAM08_b	PAM08_c	PAM08_d	PAM09	PAM10	PAM11	PAM12	PAM13	PAM14	PPL101	PPL102	PPL103	PPL104	PPL105	PPL106	PPL201	PPL202
MB compartments	y5	y5	y5B'2a	y5B'2a	B'2a	B2B'2a	B2	B2	B'2p	B'2m	B'2m	y4 <y1y2< td=""><td>y4</td><td>y4</td><td>y4</td><td>y4</td><td>B1ped</td><td>B1</td><td>a1</td><td>y3</td><td>B'1ap</td><td>B'1m</td><td>y1pedc</td><td>y1</td><td>y2a'1</td><td>a'3</td><td>a'2a2</td><td>a3</td><td>calyx</td><td>calyx</td></y1y2<>	y4	y4	y4	y4	B1ped	B1	a1	y3	B'1ap	B'1m	y1pedc	y1	y2a'1	a'3	a'2a2	a3	calyx	calyx
Total number of connectiosn from SMP108	14	19	12	4	166	4	16	9	49	153	7	5	63	1	37	10	27	17	11	1	5	3	2	5	3	2	1	4	11	5
number of cells / DAN cell type	12	7	1	2	8	4	15	1	10	11	4	5	8	5	4	1	6	6	7	11	7	8	1	1	1	1	1	1	1	1
number of connection from SMP108 / cell	1.2	2.7	12.0	2.0	20.8	1.0	1.1	9.0	4.9	13.9	1.8	1.0	7.9	0.2	9.3	10.0	4.5	2.8	1.6	0.1	0.7	0.4	2.0	5.0	3.0	2.0	1.0	4.0	11.0	5.0

Figure 4-figure supplement 2



Figure 4-figure supplement 3



Figure 5





Figure 6-figure supplement 1



Figure 7



Figure 7-figure supplement 1





Figure 8-figure supplement 1 Expression patterns of drivers