

1 **Engram reactivation during memory retrieval predicts long-term memory performance**
2 **in aged mice**

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29 **Abstract**

30 Age-related cognitive decline preferentially targets long-lasting episodic memories that
31 require intact hippocampal function. Memory traces (or engrams) are believed to be encoded
32 within the neurons activated during learning (neuronal ensembles), and recalled by
33 reactivation of the same population. However, whether engram reactivation dictates memory
34 performance late in life is not known. Here, we labelled neuronal ensembles formed during
35 object location recognition learning in the dentate gyrus, and analyzed the reactivation of this
36 population by long-term memory recall in young adult, cognitively impaired- and unimpaired-
37 aged mice. We found that reactivation of memory-encoding neuronal ensembles at long-term
38 memory recall was disrupted in impaired but not unimpaired-aged mice. Furthermore, we
39 showed that the memory performance in the aged population correlated with the degree of
40 engram reactivation at long-term memory recall. Overall, our data implicates recall-induced
41 engram reactivation as a prediction factor of memory performance in aging. Moreover, our
42 findings suggest impairments in neuronal ensemble stabilization and/or reactivation as an
43 underlying mechanism in age-dependent cognitive decline.

44

45 **Keywords:** aging, dentate gyrus, hippocampus, memory trace, neuronal ensembles, object
46 location memory.

47

48 **1. Introduction**

49 Age-related cognitive decline refers to the gradual decrease in cognitive performance
50 throughout the aging process, mostly affecting long-term storage of episodic and spatial
51 memories that depend on hippocampal function (Burke and Barnes, 2006). The dentate
52 gyrus (DG) is the hippocampal subregion most sensitive to the effects of advanced age
53 (Small et al., 2011, 2004). The DG undergoes anatomical and physiological changes as well
54 as alterations in the transcriptomic profile that are thought to underlie the aging-associated
55 dysfunction (Burke and Barnes, 2010; Janov et al., 2016). It has recently been shown that
56 subsets of DG neurons activated during learning (i.e., neuronal ensembles) belong to the
57 memory engram, and that the reactivation of this neuronal population at memory recall is
58 necessary and sufficient to evoke memory retrieval in young adult mice (Josselyn et al.,
59 2015; Tonegawa et al., 2015). Moreover, dysfunctions in neuronal ensemble reactivation and
60 memory retrieval mechanisms have been proposed as an underlying cause of dementia in
61 pathological conditions, such as Alzheimer's disease (Perusini et al., 2017; Poll et al., 2020;
62 Roy et al., 2016), in which aging presents a major risk factor. However, whether memory
63 impairments observed late in life are associated with disruptions in neuronal ensemble
64 reactivation is not known.

65 Here, we hypothesized that differences in DG neuronal ensemble dynamics underlie the
66 inter-individual variability in long-term memory (LTM) performance in the aged population.
67 We used tagging tools to label and characterize the formation and reactivation of neuronal
68 ensembles associated with long-term spatial recognition memory, a form of memory
69 particularly susceptible to decline with age. We showed that impairments in long-term spatial
70 memory during aging are associated with impaired reactivation of memory-encoding
71 neuronal ensembles at LTM recall. Moreover, we found that the degree of engram
72 reactivation in aged mice correlated with long-term spatial memory performance. Overall, our
73 findings identified novel cellular correlates of age-related memory decline.

74

75 **2. Materials and Methods**

76 2.1. *Animals*. 8-10 weeks or 18-20 months old male C57BL/6JRj mice (Janvier Labs,
77 Germany) were individually housed on a 12 h light/dark cycle and *ad libitum* access to food
78 and water (after stereotaxic surgery regular food was replaced by doxycycline-containing diet
79 (40mg/kg, BioServ, Flemington, NJ, USA)). Experiments were carried out during the light
80 phase. All of the animal procedures were approved and performed according to the
81 European Community Council Directive 86/609/EEC.

82 2.2. *Recombinant adeno associated viruses (rAAVs)*. rAAVs were produced and purified as
83 described previously (Gulmez Karaca et al., 2020; Zhang et al., 2007). The RAM-HA-GFP
84 viral construct was generated by insertion of a HA-tagged GFP expression cassette into the
85 multiple cloning site (MCS) of rAAV-RAM (rAAV-RAM-d2TTA::TRE-MCS-WPRE-pA) that
86 was a kind gift from Yingxi Lin (Addgene plasmid # 63931; <http://n2t.net/addgene:63931>;
87 RRID:Addgene_63931)) (Gulmez Karaca et al., 2020; Sørensen et al., 2016).

88 2.3. *Stereotaxic surgery*. 250 nl of rAAV-RAM viral solution was injected into the dentate
89 gyrus (DG) of the hippocampus at the following coordinates relative to Bregma: – 2.0 mm
90 anteroposterior, ± 1.3 mm medio-lateral, – 2.4 mm dorsoventral at the rate of 80-100 nl / min.
91 The needle was left in place for 10 min before and after each injection to allow DG-specific
92 diffusion of the fluid. Only the mice with viral spread throughout the DG of at least 60 µm
93 range in the AP axis were considered for behavior and further analysis. Throughout the
94 study, a total of 10 mice were rejected from the analysis due to insufficient viral spread (5
95 mice) and/or lack of exploration during behavioral training or testing (5 mice).

96 2.4. *Behavior paradigms*. Three weeks after stereotaxic surgery and following 5 days of
97 handling, the spatial object recognition (SOR) test was performed. SOR training involved four
98 sessions; each lasted 6 min and was separated by 3 min intervals in the home-cage. In the
99 first session, mice explored an empty black, square open field (50 cm × 50 cm × 50 cm) with
100 a visual cue placed on one wall. This session was also used for the open field test as
101 described previously (Brito et al., 2020b, 2020a; Gulmez Karaca et al., 2018). The Smart
102 Video Tracking Software (Panlab, Harvard Apparatus) was used to score the time spent in
103 the central zone (32% of the arena), number of center entries and total distance travelled in

104 the open field test. In the next three sessions of the SOR training, mice freely explored two
105 diagonally located distinct objects (a glass bottle and a metal tower) in the arena. After the
106 training, mice were placed in the home-cage and were undisturbed until the memory test.
107 During the memory test (24 h post-training), one of the two objects was displaced in the
108 arena (while the other one remained in the original position), which the mice explored freely
109 for 6 min. Object exploration was defined as the animal sniffing the object or pointing its nose
110 to the object within 1-cm distance. When the animal leaned on the object but did not direct
111 the nose towards it, object exploration was not considered. SOR memory performance was
112 assessed by the formula ($T_{\text{displaced}} / (T_{\text{nondisplaced}} + T_{\text{displaced}}) \times 100$) where T represents time
113 exploring an object. Assigning the aged mice into cognitively impaired and unimpaired
114 groups was performed based on the mean SOR memory performance of the young group.
115 Aged mice displaying memory performance lower than one standard deviation from the
116 mean of young mice were designated as cognitively impaired.

117 *2.5. Immunohistochemistry.* 2 h after the start of the SOR test, mice were perfused
118 intracardially with 4% paraformaldehyde (PFA) (Sigma, Munich, Germany) and free-floating
119 brain slices (at a thickness of 20 μm) were immunostained as previously described (Gulmez
120 Karaca et al., 2018; Oliveira et al., 2012). Primary antibodies were used at the following
121 concentrations: anti-HA tag (Covance MMS-101R (1:1000)), anti-Fos (Cell Signaling #2250,
122 1:1000)).

123 *2.6. Image acquisition and analysis.* Z-stacks of DG images (3 frames with 2 μm interval at
124 20x magnification) were acquired with Nikon A1R confocal microscope (at Nikon Imaging
125 Center, BioQuant, Heidelberg) using NIS-Elements software or with Leica SP8 confocal
126 microscope using LAS X software. Maximum projection files of each stack were imported in
127 Fiji (Schindelin et al., 2012), and Fos⁺ or HA-GFP⁺ neurons were manually marked after
128 background subtraction and application of a signal threshold. Reactivation rate
129 $((\text{GFP}^+\text{Fos}^+)/(\text{GFP}^+)\times 100)$, and similarity index $((\text{GFP}^+\text{Fos}^+)/((\text{GFP}^+)+(\text{Fos}^+)-$
130 $(\text{GFP}^+\text{Fos}^+))\times 100$ was calculated as previously described (Cowansage et al., 2014; Gulmez
131 Karaca et al., 2020; Milczarek et al., 2018). Data was normalized to the mean of the aged

132 impaired group to avoid artifacts that may be caused from different viral expressions among
133 experimental batches.

134 *2.7. Statistical analysis.* Blinding to experimental conditions was applied to image and
135 behavioral analysis. Each data was subjected to a normality test (Shapiro-Wilk normality test,
136 $\alpha=0.05$) before further analysis. For normally distributed data ordinary one-way ANOVA
137 followed by Tukey's multiple comparisons test, whereas for non-normally distributed data,
138 Kruskal-Wallis test followed by Dunn's multiple comparisons test was used to compare three
139 groups. When comparing two dependent data sets, paired t-test was performed if the data
140 set showed a normal distribution, if not, Wilcoxon matched-pairs signed rank test was
141 applied. For correlation analysis, Pearson correlation test was applied when the data
142 followed a normal distribution. Nonparametric Spearman correlation was computed in case of
143 a non-normally distributed dataset. Statistical analysis was performed using GraphPad prism
144 for Mac OS X, version 8.

145

146 **3. Results**

147 *3.1. Neuronal ensemble tagging in the DG of young, cognitively-impaired or -unimpaired*
148 *aged mice.* We aimed at characterizing DG neuronal ensembles that hold long-term
149 representations of object location memory in order to identify possible correlates of cognitive
150 dysfunction in the aged population. To label the neuronal ensemble activated during learning,
151 we used the robust activity marking (RAM) system (Sørensen et al., 2016) that allowed
152 learning-dependent expression of HA-tagged GFP (Sørensen et al., 2016) (**Figure 1A**). We
153 have previously confirmed that this tool reliably tags the DG neuronal ensembles associated
154 with object location memory (Gulmez Karaca et al., 2020). First, we stereotaxically delivered
155 recombinant adeno associated viruses (rAAVs) containing the RAM-HA-GFP viral construct
156 into the DG of young-adult (2 months old) or aged (18-20 months old) mice. Through the
157 removal of doxycycline diet, we tagged the neuronal population activated during object
158 location training (**Figure 1A**). When tested 24 h after training all young adult mice displayed
159 higher than chance (50%) preference for the displaced object (**Figure 1B**), indicating intact

160 long-term object location memory. In contrast, the performance of the aged group was
161 heterogenous; some mice performed similarly to young mice, whereas others showed
162 preferences closer to chance level (**Figure 1B**). To be able to characterize the aged
163 population according to individual differences in cognitive performance, we sorted the aged
164 group into aged impaired (AI) and aged unimpaired (AU) based on the mice's long-term
165 spatial memory scores (**Figure 1B**). Aged mice displaying preferences for the displaced
166 object lower than one standard deviation below the mean of young mice (64.3%, represented
167 with a dashed line in Figure 1B) were considered impaired. Importantly, we confirmed that all
168 groups displayed comparable object exploration times during the training session (**Figure**
169 **1C**) and demonstrated similar locomotion and anxiety-like behavior in open-field test (**Figure**
170 **1D-F**). This indicates that poor LTM performance in AI mice was not due to lower exploratory
171 behavior or motor or anxiety impairments.

172 *3.2. Cognitively impaired aged mice exhibit impaired recall-induced neuronal ensemble*
173 *reactivation in the upper blade of the DG.* Next, we investigated the reactivation of spatial
174 memory-associated neuronal ensembles in response to LTM recall. To this end, the
175 expression of HA-GFP and Fos in the DG, that represents active neurons during learning or
176 LTM recall respectively, was analyzed 2 h after object location test (**Figure 2A**). The
177 neuronal population coexpressing HA-GFP and Fos represented the population active in both
178 episodes, i.e., reactivated neuronal ensemble (**Figure 2B**). Given the previously described
179 differences in responsiveness of the upper and lower blades of the DG to environmental
180 exposure (Chawla et al., 2013; Gulmez Karaca et al., 2020; Marrone et al., 2012; Ramirez-
181 Amaya et al., 2013), we analyzed the two subregions separately. We found that AI mice
182 exhibited reduced recall-driven reactivation of the neuronal population activated by learning
183 in the DG compared to young or AU mice (**Figure 2C**), whereas the reactivation rates
184 observed in the DG of young and AU mice were similar (**Figure 2C**). Interestingly, the
185 impaired reactivation rate was specific for the neuronal ensembles located in the upper, but
186 not the lower, blade of the DG in AI mice (**Figure 2C**). Although AI mice were able to
187 reactivate the learning-activated neuronal population at a rate significantly higher than what

188 would occur by chance, the observed overlap above chance level was higher both in the
189 young and AU groups compared to the AI group (**Figure 2D**). The analysis of the two DG
190 blades independently, showed that, similarly to what we observed for the reactivation rate
191 (**Figure 2C**), the effect originated from the upper blade (**Figure 2D**). There was no significant
192 difference in the reactivation above chance levels of young and AU mice (**Figure 2D**).
193 Furthermore, to evaluate whether the learning-activated neuronal ensemble pattern was
194 reinstated equally in all of the three groups at memory recall, we applied a second formula,
195 the similarity index (Milczarek et al., 2018). This formula complements reactivation rate by
196 also including the number of the neurons activated solely by recall, and thereby allows to
197 measure the degree of similarity in the neuronal activity pattern at learning and recall. We
198 observed that AI mice exhibited the same selective impairment in the DG upper blade
199 (**Figure 2E**). We then confirmed that the identified differences between the AI and AU were
200 not due to differences in the number of neurons activated by learning (GFP⁺) (**Figure 2F**) or
201 by recall (Fos⁺) (**Figure 2G**) in the aged population. Although learning triggered the activation
202 of fewer neurons in the AU mice than the young (**Figure 2F**), this did not correlate with
203 reactivation rate in the DG upper blade ($p=0.2286$, $r=-0.3073$, Spearman correlation, data not
204 shown) or with memory performance ($p=0.1566$, $r=-0.3594$, Pearson correlation, data not
205 shown) in the aged. Altogether, this set of experiments revealed that despite being able to
206 activate a similar proportion of neurons upon learning or memory recall, cognitively impaired
207 mice exhibit disrupted neuronal ensemble reactivation at LTM recall in the upper blade of the
208 DG.

209 *3.3. LTM performance of aged mice correlates with the reactivation rate in the upper blade of*
210 *the DG.* Finally, we tested whether the neuronal ensemble reactivation rate in the upper DG
211 blade correlates with memory performance in aged mice. For this, we performed Pearson
212 correlation analysis between LTM performance and reactivation rate in the aged population
213 (i.e., pooled data of AI and AU mice). Remarkably, we found a significant positive correlation
214 between the rate of neuronal ensemble reactivation in the total DG and the memory score of
215 the mice ($p=0.0272$, $r=0.5396$, Spearman correlation, data not shown) which primarily

216 originated from the upper blade of the DG (**Figure 2H**), but was not present in the lower
217 blade of the DG ($p=0.3668$, $r=0.2496$, Spearman correlation, data not shown). Similar to the
218 reactivation rate, we observed a correlation between the degree of the similarity index in the
219 total DG and long-term memory performance in aged mice ($p=0.0064$, $r=0.6437$, Spearman
220 correlation, data not shown) that was selectively present in the upper blade of the DG
221 (**Figure 2I**), but not in the lower blade of the DG ($p=0.6874$, $r=0.1128$, Spearman correlation,
222 data not shown). Overall, these results demonstrated that cognitive performance during
223 aging positively correlates with neuronal ensemble reactivation during the recall of LTM.

224

225 **4. Discussion**

226 In this study, we tagged the neuronal ensemble formed during a spatial recognition task in
227 the DG of young, aged-impaired and aged-unimpaired mice and identified a cellular correlate
228 of cognitive performance in aged mice. We showed that cognitively impaired aged mice
229 exhibit impairments in recruiting the original memory engram at long-term spatial memory
230 recall compared to young or cognitively unimpaired aged mice. We further showed that the
231 degree of engram reactivation correlates with the strength of LTM in the aged population.
232 We focused on long-term object location memory which has previously been shown to be
233 impaired by aging (Oliveira et al., 2012; Wimmer et al., 2012). Increasing evidence suggests
234 that each memory is encoded by a subset of neurons (neuronal ensembles) that are
235 synchronously activated upon learning and reactivated by the retrieval of the memory
236 (Josselyn et al., 2015; Tonegawa et al., 2015). A previous study by Penner and colleagues
237 showed that neuronal reactivation upon short-term re-exposure to a previously visited
238 context is reduced in the DG, but not CA1, of aged compared to young adult mice (Penner et
239 al., 2011). These findings suggest that impairments in neuronal reactivation may be
240 associated with cognitive deficits in aged individuals (Penner et al., 2011). Here, we
241 characterized DG neuronal ensemble during the formation and retrieval of long-term
242 recognition memory and correlated with the performance in the same task. We found that
243 aging-related impairments in long-term object location memory are not associated with the

244 size of the DG neuronal population activated by learning or recall, but rather with the fidelity
245 of reactivation of the encoding population during memory retrieval. Aged impaired and
246 unimpaired mice showed similar number of neurons activated by learning or by recall. In
247 contrast, cognitively impaired mice had significantly disrupted reactivation rates compared to
248 unimpaired aged, or young adult mice. Intriguingly, the differences were found in the
249 neuronal population located in the upper, but not lower, blade of the DG. This is in line with
250 previous studies showing that behaviorally-induced expression of activity regulated genes
251 occurs primarily in the upper blade of the DG (Chawla et al., 2013, 2005; Erwin et al., 2020;
252 Gulmez Karaca et al., 2020; Marrone et al., 2012; Ramirez-Amaya et al., 2013). These
253 reports indicate that the sparse population of DG upper blade neurons form the spatial
254 memory neuronal ensemble. Therefore, it may be expected that alterations in ensemble
255 dynamics are specific to this subregion. Finally, our findings showed a positive correlation
256 between the neuronal ensemble reactivation rate in the upper blade of the DG and LTM
257 performance of aged mice. These findings suggest a novel mechanism that may underlie
258 long-term spatial memory integrity in the aged population.

259 The underlying molecular and physiological causes of disrupted neuronal ensemble
260 reactivation in aged cognitively-impaired aged mice are not understood. It is well established
261 that aging is accompanied by anatomical and physiological changes in the DG. Namely,
262 fewer synaptic contacts and impaired synaptic plasticity in aged versus young adults have
263 been reported (Burke and Barnes, 2010). Interestingly, age-related impairments in LTP
264 maintenance correlated with memory performance (Rosenzweig and Barnes, 2003). Thus,
265 deficits in the reactivation of the original encoding neuronal population may result from
266 impaired neuronal ensemble stabilization during memory consolidation, which could emerge
267 from the inability to strengthen the connectivity between ensemble neurons.

268 Recently, a few studies showed that LTM impairments in a mouse model of Alzheimer's
269 disease (AD) are linked to deficits in neuronal ensemble reactivation (Perusini et al., 2017;
270 Poll et al., 2020; Roy et al., 2016). Our findings now show that alterations in neuronal
271 ensemble properties are a common mechanism in aging and aging-associated pathological

272 conditions. This underscores the need for therapeutic approaches targeted at facilitating
273 engram reactivation to restore age-associated memory impairments.

274

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281

282 **Disclosure statement**

283 The authors declare no competing financial interests.

284

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369

370 **Figure legends**

371

372 **Figure 1.** Tagging neuronal ensembles formed during spatial object-location learning in the DG of
373 young, cognitively-impaired or -unimpaired aged mice. A) Schematic representation of the RAM-based
374 viral construct used for neuronal ensemble tagging and of the experimental design. B) Percentage of
375 preference for displaced object during the object location test session. Dashed line (64.3%) represents
376 the threshold applied to categorize AI (n=9) and AU (n=8) mice (Y: mean=76.07, SD=11.80, n=16). C)
377 Total object exploration time during the training session of object-location test ($p=0.9237$, Kruskal-
378 Wallis test with Dunn's multiple comparisons: Y (n=13) vs AU (n=8): $p>0.9999$; Y vs AI (n=9):
379 $p>0.9999$; AU vs AI: $p>0.9999$). D) Representative trajectories during open-field test. E) Total
380 distance travelled in open-field test ($F(2, 27)=0.0072$, $p=0.9303$, One-way ANOVA with Tukey's

381 multiple comparisons test: Y (n=13) vs AU (n=8): $p=0.9694$; Y vs AI (n=9): $p=0.9287$; AU vs AI:
382 $p=0.9936$). F) Percentage of time in the center ($p=0.8604$, Kruskal-Wallis test with Dunn's multiple
383 comparisons: Y (n=13) vs AU (n=8): $p>0.9999$; Y vs AI (n=9): $p>0.9999$; AU vs AI: $p>0.9999$) and
384 number of center entries ($F(2,27)=0.05143$, $p=0.9500$, one-way ANOVA with Tukey's multiple
385 comparisons: Y (n=13) vs AU (n=8): $p=0.9826$; Y vs AI (n=9): $p=0.9468$; AU vs AI: $p=0.9926$) in open-
386 field test. ns: not significant. ITR: inverted terminal repeat, DG: Dentate gyrus of the hippocampus,
387 Dox: doxycycline, AU: aged-unimpaired, AI: aged-impaired, Y: young, rAAV: recombinant adeno-
388 associated viruses, TRE: tetracycline responsive element, pRAM: robust activity marking promoter.
389 Error bars represent s.e.m.

390

391 **Figure 2.** Age-related cognitive decline is associated with disrupted DG neuronal ensemble
392 reactivation at LTM recall. A) Schematic representation of the experimental design. B) Representative
393 images showing the immunohistochemical analysis of HA-GFP and Fos expression in young, AU and
394 AI groups. Scale bars represent 50 μm . Dashed lines define DG granule cell layer. White arrows
395 indicate cells with overlapping HA-GFP and Fos signal, whereas cyan and magenta arrows indicate
396 individual HA-GFP and Fos signals, respectively. C) Normalized reactivation rates of learning-
397 activated neuronal ensembles at memory recall (Total DG: $p=0.0690$, Kruskal-Wallis test with Dunn's
398 multiple comparisons: Y vs AU: $p>0.9999$; Y vs AI: $p=0.1484$; AU vs AI: $p=0.1146$; Upper blade:
399 $p=0.0003$, Kruskal-Wallis test with Dunn's multiple comparisons: Y vs AU: $p>0.9999$; Y vs AI:
400 $p=0.0009$; AU vs AI: $p=0.0016$; Lower blade: $p=0.8702$, Kruskal-Wallis test with Dunn's multiple
401 comparisons: Y vs AU: $p>0.9999$; Y vs AI: $p>0.9999$; AU vs AI: $p>0.9999$). D) Observed- over chance-
402 overlap between HA-GFP and Fos expressions in young, AU and AI groups (Observed- vs chance-
403 overlap: Total Y: $W = 91$, $p = 0.0002$ by Wilcoxon test; Total AU: $W = 36$, $p = 0.0078$ by Wilcoxon test;
404 Total AI: $W = 43$, $p = 0.0078$ by Wilcoxon test; Upper blade Y: $W = 91$, $p = 0.0002$ by Wilcoxon test;
405 Upper blade AU: $W = 36$, $p = 0.0078$ by Wilcoxon test; Upper blade AI: $W = 43$, $p = 0.0078$ by
406 Wilcoxon test; Lower blade Y: $W = 89$, $p = 0.0005$ by Wilcoxon test; Lower blade AU: $W = 34$, $p =$
407 0.0156 by Wilcoxon test; Lower blade AI: $W = 33$, $p = 0.0547$ by Wilcoxon test) (Observed/chance:
408 Total DG: $F(2,27)=7.201$, $p=0.0031$, one-way ANOVA with Tukey's multiple comparisons: Y vs AU:
409 $p=0.6181$; Y vs AI: $p=0.0152$; AU vs AI: $p=0.0039$; Upper blade: $p=0.0010$, Kruskal-Wallis test with
410 Dunn's multiple comparisons: Y vs AU: $p>0.9999$; Y vs AI: $p=0.0062$; AU vs AI: $p=0.0022$; Lower
411 blade: $F(2,27)=1.336$, $p=0.2797$, one-way ANOVA with Tukey's multiple comparisons: Y vs AU:

412 $p=0.5108$; Y vs AI: $p=0.7924$; AU vs AI: $p=0.2557$). E) Normalized similarity indices of neuronal
413 ensemble populations activated at learning or at LTM recall (Total DG: $p=0.0299$, Kruskal-Wallis test
414 with Dunn's multiple comparisons: Y vs AU: $p>0.9999$; Y vs AI: $p=0.0493$; AU vs AI: $p=0.0789$; Upper
415 blade: $p=0.0002$, Kruskal-Wallis test with Dunn's multiple comparisons: Y vs AU: $p>0.9999$; Y vs AI:
416 $p=0.0003$; AU vs AI: $p=0.0050$; Lower blade: $p=0.6309$, Kruskal-Wallis test with Dunn's multiple
417 comparisons: Y vs AU: $p>0.9999$; Y vs AI: $p>0.9999$; AU vs AI: $p>0.9999$). F) Percentage of HA-GFP⁺
418 neurons (Total DG: $p=0.0275$, Kruskal-Wallis test with Dunn's multiple comparisons: Y vs AU:
419 $p=0.0395$; Y vs AI: $p=0.1667$; AU vs AI: $p>0.9999$; Upper blade: $p=0.0123$, Kruskal-Wallis test with
420 Dunn's multiple comparisons: Y vs AU: $p=0.0160$; Y vs AI: $p=0.1255$; AU vs AI: $p>0.9999$; Lower
421 blade: $p=0.0842$, Kruskal-Wallis test with Dunn's multiple comparisons: Y vs AU: $p=0.1364$; Y vs AI:
422 $p=0.2767$; AU vs AI: $p>0.9999$). G) Percentage of Fos⁺ neurons (Total DG: $p=0.1560$, Kruskal-Wallis
423 test with Dunn's multiple comparisons: Y vs AU: $p=0.4933$; Y vs AI: $p=0.2275$; AU vs AI: $p>0.9999$;
424 Upper blade: $p=0.1724$, Kruskal-Wallis test with Dunn's multiple comparisons: Y vs AU: $p=0.4823$; Y
425 vs AI: $p=0.2690$; AU vs AI: $p>0.9999$; Lower blade: $F(2,27)=3.460$, $p=0.0459$, one-way ANOVA with
426 Tukey's multiple comparisons: Y vs AU: $p=0.0665$; Y vs AI: $p=0.1263$; AU vs AI: $p=0.9298$). H)
427 Spearman correlation between the normalized neuronal ensemble reactivation in the upper blade of
428 the DG and memory performance of aged mice at long-term object-location test ($p<0.0001$, $r=0.8396$,
429 $n=17$, Spearman correlation). I) Spearman correlation between the normalized similarity index in the
430 upper blade of the DG and memory performance of aged mice at long-term object-location test
431 ($p<0.0001$, $r=0.8236$, $n=17$, Spearman correlation). In all graphs: Y ($n=13$), AI ($n=9$), AU ($n=8$),
432 * $p<0.05$; ** $p<0.01$; *** $p<0.001$; # $p<0.05$; ## $p<0.01$; ### $p<0.001$; ns: not significant. DG: Dentate gyrus of
433 the hippocampus, Dox: doxycycline, AU: aged-unimpaired, AI: aged-impaired, rAAV: recombinant
434 adeno-associated viruses, pRAM: robust activity marking promoter. Error bars represent s.e.m.
435

Figure 1**A**

Stereotaxic delivery of rAAVs into the DG

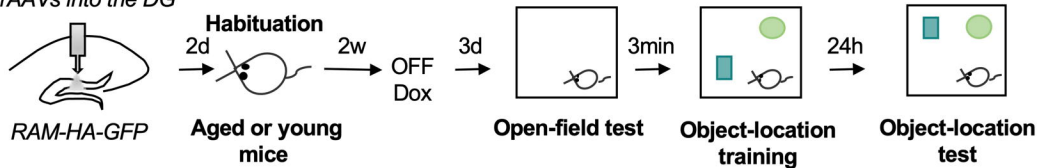
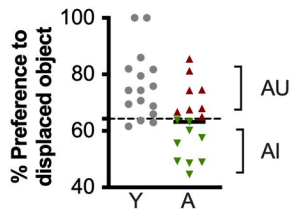
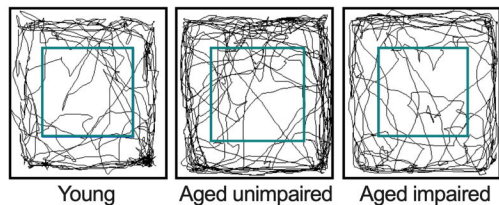
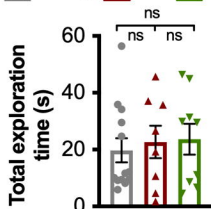
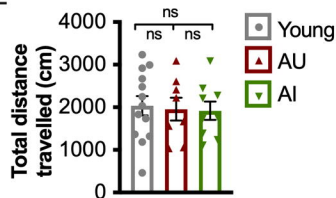
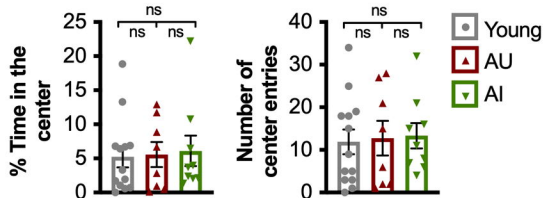
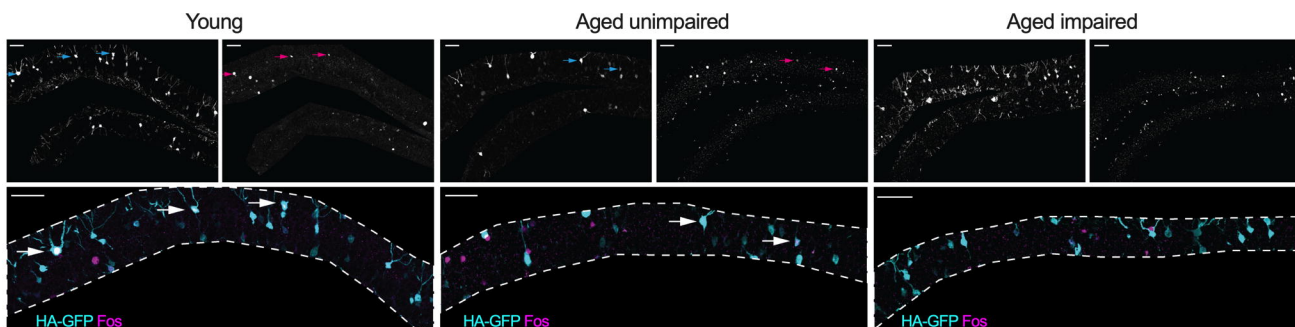
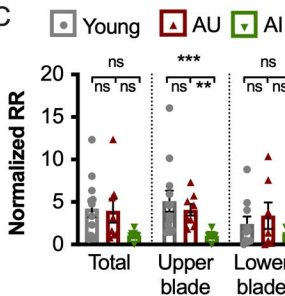
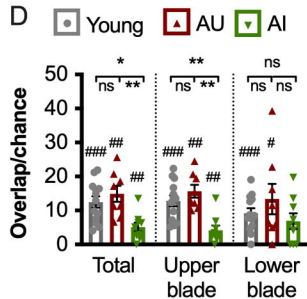
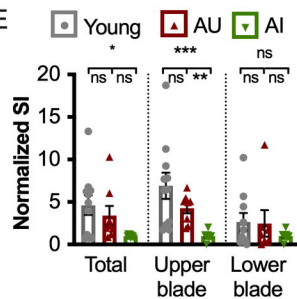
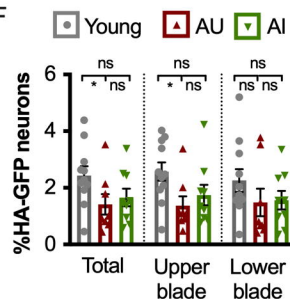
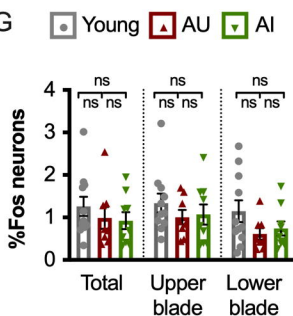
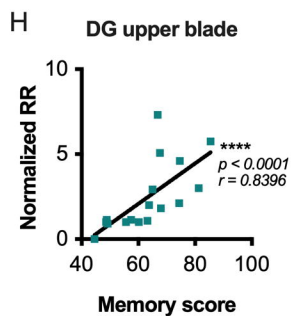
**B****C** **D****E****F**

Figure 2**A**

Stereotaxic delivery
of rAAVs into the DG

**B****C****D****E****F****G****H****I**