Title

Early spatial and contextual coding deficits in hippocampal CA1 precede performance

decline in an Alzheimer's disease model

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Abstract

Alzheimer's disease (AD) is characterised by progressive memory decline, yet how hippocampal representations change at early disease stages remains largely unknown. Using a real-world head-fixed spatial alternation task combined with in vivo two-photon imaging, we examined hippocampal CA1 representations in 5xFAD mice. Although 7-12 month old 5xFAD mice exhibited significant impairments in task learning, 2-4 month old mice performed normally, allowing us to assess hippocampal coding before behavioural decline. At this early stage, CA1 neurons showed intact global spatial encoding but impaired task-related place cell representations. Contextual representations revealed by trajectory-dependent coding were weakened at both the population and single cell levels, accompanied by a shift from predominantly retrospective towards more prospective encoding along the central arm. Despite these vulnerabilities, a subset of place cells largely retained stable spatial representations that persisted throughout task learning in early-stage AD mice. Chemogenetic activation of basal forebrain cholinergic neurons improved spatial information rates and stability of spatial and contextual coding in early-stage 5xFAD mice. Together, these findings show that hippocampal spatial and contextual coding is disrupted early in AD, while a subset of stable place cells remains preserved, and that these representational deficits emerge before impairments in spatial alternation behaviour.

Introduction

Alzheimer's disease (AD) is the most prevalent cause of dementia in older adults, particularly affecting individuals over 65 years of age [1]. A central feature of AD pathology is the amyloidogenic processing of amyloid precursor protein (APP), which produces amyloid-β (Aβ) and triggers downstream pathophysiological cascades leading to synaptic dysfunction and neuronal loss [2, 3, 4, 5, 6]. Among the cognitive impairments associated with this pathway, spatial navigation deficits are especially prominent and often emerging as early as the mild cognitive impairment stage [7, 8]. More than 60% of AD patients experience spatial disorientation and wandering behaviour [9]. These deficits are closely linked to hippocampal dysfunction. Hippocampal CA1 neurons, crucial for forming spatial and episodic memories, exhibit abnormalities in morphology, synaptic plasticity, and neurogenesis in AD, in both human patients and mouse models [10, 11, 12].

Hippocampal place cells are essential for supporting spatial navigation and memory formation, and accumulating evidence indicates that their function is disrupted in AD. Impairments in place cell firing have been reported across several AD mouse models, including APP transgenic and APP knock-in lines. For example, 11-16 month age Tg2576 and APP mice show degraded place fields and reduced spatial information, typically in conjunction with behavioural deficits [13, 14]. In 3xTg AD mice, reduced spatial information has been reported in young 3xTg mice during open-field or linear track running , with reduced theta- and slow-gamma phase locking [15, 16]. In 5xFAD mice, impaired synaptic efficacy and weakened sharp wave-ripples have been observed during virtual reality

spatial tasks at 11–14 months of age[17]. Object-discrimination experiments in 4 month-old

5xFAD mice revealed preserved task performance and population coding within the

discrimination task, despite reduced spatial coding in other environments [18]. Overall,

these findings demonstrate place cell dysfunction in AD, though most evidence comes

from later disease stages or from simplified exploratory paradigms rather than

behaviourally demanding tasks.

In contrast to place cells, which encode spatial location, splitter cells encode trajectory-

dependent contextual information, firing differently on a common path depending on past

choices or future decisions. First identified in continuous alternation tasks in rats [19, 20],

splitter cells integrate retrospective and prospective information within single CA1

neurons and are thought to support flexible navigation [21]. Despite their importance, little

is known about how splitter cell activity is affected in AD. Given that Aβ-related synaptic

changes impair hippocampal learning [22] and that splitter cell activity depends on rapid

synaptic plasticity [23], trajectory-dependent contextual coding may be particularly

vulnerable early in the disease.

Here, we developed a real-world head-fixed continuous T-maze alternation task that

imposes goal-directed spatial learning demands comparable to those in freely moving

paradigms[24, 25]. We combined this task with two-photon calcium imaging in 5xFAD mice,

which begin to accumulate amyloid pathology around 2-4 months of age due to APP and

presenilin-1 mutations[26, 27], allowing us to record CA1 neuronal activity during task

performance. This approach enabled us to examine how spatial coding and trajectory-

dependent contextual representations in CA1 are altered during the early stage of AD.

Given the established role of basal forebrain cholinergic input in modulating hippocampal

synaptic plasticity[28], we further tested whether enhancing cholinergic activity could

improve CA1 encoding in early-stage 5xFAD mice.

Results

1. A head-fixed alternation task reveals preserved spatial learning in 2- 4 month old

5xFAD mice and emerging deficits with age

We developed a novel behavioural paradigm to assess spatial memory in head-fixed mice,

in which animals performed a continuous spatial alternation task within a real-world, air-

lifted Y-maze (Fig. 1a, left). Mice received a 4 µl of liquid reward only when they entered

the arm opposite to the one visited in the previous trial (Fig. 1a, right). Water-restricted

mice expressing GCaMP7s in CA1 pyramidal neurons of the dorsal hippocampus were

trained to perform the task while calcium activity was simultaneously recorded using two-

photon microscope (Fig. 1b).

Consistent with previous reports, amyloid plaques were observed in 5xFAD mice at 2-4

months ages and became extensive in the hippocampal CA1 region at 7-12 months [29, 30].

(Supplementary Fig. 1a). We therefore focused on these two age groups to examine task

performance. In 2-4 month-old 5xFAD mice, the percentage of correct trials was

comparable to age-matched controls (Genotype × Day, β = -1.29, 95%CI [-2.63, 0.44], p

= 0.16, linear-mixed effect mode (LMM); Fig. 1c,d). In contrast, 7-12 month-old 5xFAD

mice were exhibited pronounced learning impairments (Genotype × Day, β = -2.56,

95%CI [-4.57, -0.54], p = 0.013, Fig. 1f,g). Notably, aged 5xFAD mice also showed greater

variability in the number of trials completed per day, potentially affecting learning

opportunities (Supplementary Fig. 1b,c,f,g). To account for this, trials from all sessions

were grouped into consecutive 50-trial blocks, and the percentage of correct trials was

calculated per block (Supplementary Fig. 1d, h). The analysis yielded results consistent

with day wise performance: 2-4 month-old 5xFAD mice displayed no significant

behavioural deficits (Genotype × Block, β = 0.35, 95%CI [-0.70, 1.40], p = 0.51; Fig. 1e),

whereas 7-12 month-old mice showed a significant impairment (Genotype × Block, β = -

1.85, 95%CI [-3.63, -0.06], p = 0.043; Fig. 1h).

Furthermore, the proportion of 7-12 month-old 5xFAD mice reaching the task acquisition

criterion was significantly reduced compared to controls, whereas the 2-4 month-old

group showed normal acquisition rates (Supplementary Fig. 1e,i). These findings indicate

that spatial learning deficits are evident in 5xFAD mice by 7-12 months of age, consistent

with previous studies [27], while 2-4 month-old 5xFAD mice retain intact task performance.

To evaluate the overall spatial coding capacity of CA1 neurons, we quantified global

spatial information, defined as the Skaggs' spatial information content computed from

two-dimensional spatial activity maps across all trajectories within the imaging session.

In 2–4 month-old 5xFAD mice, CA1 neurons exhibited spatial information levels

comparable to controls (β = -0.12, 95% CI [-0.89, 0.66], p = 0.77; LMM; Fig. 1i, left).

Similarly, the proportion of neurons meeting the place cell criterion showed a non-

significant reduction (rrb = -0.25, p = 0.25; Fig. 1i, right). In contrast, 7-12 month-old

5xFAD mice displayed significant impairments in spatial coding, as reflected by reduced

global spatial information (β = -0.54, 95% CI [-1.00, -0.09], p = 0.019; Fig. 1j, left) and a

lower proportion of place cells (rrb = -0.67, p = 0.015; Fig. 1j, right). Together, these

results demonstrate that while 2-4 month old 5xFAD mice retain intact spatial learning

and CA1 global spatial coding during spatial alternation task, both behavioural

performance and neuronal spatial representations impaired markedly by 7-12 months of

age.

Figure 1. Impairments in behavioural deficits and CA1 spatial representations in 5xFAD

mice during the spatial alternation task

a. Left: Schematic illustration of a head-fixed mouse navigating an air-lifted figure-8 maze. Right:

Bird's eye view of the same maze, with correct trials highlighted in blue (left-to-right) and red

(right-to-left).

b. Schematic of in vivo calcium imaging from CA1 pyramidal neurons expressing jGCaMP7s. Twelve example regions of interest (ROIs) are overlaid on the imaging field of view, with corresponding calcium ΔF/F traces shown right. Scale bar: 100 μm.

c. Representative temporal trajectories of mouse navigation on early and final training days (2–4 months of age). Correct trials are shown in black, and error trials in red. Reward zones are indicated by blue dots.

d. Learning curves showing the percentage of correct alternation trials over training days in 2–4 month-old mice. The horizontal dotted line indicates the 70% performance crietria. Control (Ctrl; n = 10 mice) vs. 5xFAD (AD; n = 7 mice). Genotype × Day interaction, linear-mixed effect mode (LMM).

e. Trials across all sessions were binned into consecutive 50-trial blocks, with the percentage of correct% calculated per block (Genotype × Block interaction). Curves show quadratic mixed-effects model fits, with shaded areas indicating 95% confidence intervals.

f - h. Same as in panels **c - e**, but for mice aged 7–12 months (Ctrl: n = 6 mice; AD: n = 6 mice).

i. Left: Cumulative distribution of Skagg's spatial information for all detected CA1 neurons, based on 2D spatial maps from 2-4 month-old mice. Inset: session level medians. LMM test. Right: Proportion of neurons classified as place cells per session. Two-sided Wilcoxon rank-sum test. Ctrl: n = 4397 cells from 20 imaging sessions (10 mice) ; AD: n = 3073 cells from 13 sessions (7 mice).

j. Same as panel **i**, for mice aged 7-12 months. Ctrl: n = 2168 cells from 12 imaging sessions (6 mice); AD: n = 1477 cells from 8 sessions (6 mice).

2. Disruption of CA1 place cell coding during task-related trajectories in 2-4 month old 5xFAD mice.

Understanding the early stage of AD is crucial for identifying initial circuit level disruptions and informing early intervention [31, 32]. Based on our behavioural and neuronal observations (Fig. 1c-j), together with prior reports indicating that hippocampal long-term potentiation and detectable hippocampal-dependent behavioural deficits emerge after 6 months of age in 5xFAD mice [29, 33, 34], we focused on 2-4 month old 5xFAD mice to examine early alternations in hippocampal neural coding.

To assess task-related spatial representations, we linearised trajectories and generated one-dimensional spatial activity maps during the post-acquisition phase, when mice consistently performed the alternation task. Analyses were restricted to correct trials, which most directly reflect task engagement. This yielded a measure of spatial coding specifically aligned with behavioural performance, in contrast to global spatial metrics. At the population-level, spatial stability was quantified by calculating population vector (PV) correlations between odd and even trials across all CA1 neurons within each imaging session (Fig. 2a). 2-4 month old 5xFAD mice exhibited significantly reduced PV correlations compared to controls (rrb = -0.45, p = 0.031; Fig. 2b), indicating impaired population level spatial consistency despite intact behavioural performance (Supplementary Fig. 2a-e).

We next examined spatial coding stability at the single cell level by analysing CA1 place cell activity during correct trials (Fig. 2c). In 5xFAD mice, place cells showed a trend

toward reduced spatial information relative to controls (β = -0.24, 95% CI [-0.48, 0.006],

p = 0.056; LMM; Fig. 2d). Trial-to-trial tuning curve correlation, a measure of place cell

stability, was significantly lower in 5xFAD mice (β = -0.02, 95% CI [-0.04, -0.002], p =

0.029; Fig. 2e,f), indicating decreased consistency of spatial representations across

repeated trajectories. To further examine spatial coding stability, we quantified (i) the

active trial ratio, defined as the proportion of trials in which a place cell was active within

its place field, and (ii) the tuning curve correlation across adjacent active trials. Both

measures were significantly reduced in 5xFAD mice (Supplementary Fig. 2f,g), further

supporting disrupted trial-wise spatial coding. Place field fidelity was assessed using two

measures: (i) in-field versus out-of-field selectivity, which showed a non-significant trend

towards reduction (β = -0.04, 95% CI [-0.08, 0.002], p = 0.064; Fig. 2g); and (ii) peak in-

field activity, which was significantly lower in 5xFAD mice (β = -0.33, 95% CI [-0.57, -0.08],

p = 0.0097; Fig. 2h). Together, these results suggest that CA1 place cells in early-stage

5xFAD mice exhibit reduced spatial encoding stability and weakened peak activity within

place fields.

Importantly, behavioural performance and locomotion were preserved during imaging

sessions (Supplementary Fig. 2a-e), confirming that neuronal deficits were not driven by

overt behavioural changes. Additionally, in both groups, place cell stability and fidelity

differed between correct and error trials (Supplementary Fig. 2h-l), indicating the task-

dependence of CA1 spatial coding. Together, these findings show that CA1 spatial coding

is already impaired in early-stage 5xFAD mice, with reduced population stability, weaker

place field activity, and less reliable single cell responses during alternation task-related

navigation.

Figure 2. Impaired CA1 place cells during correct trials in 2-4 month-old 5xFAD mice.

a. Population vector (PV) correlations between odd and even trials for all detected CA1 neurons,

averaged across animals within each group.

b. Left: Diagonal PV correlation coefficients plotted along the linearised trajectory. Right: Mean

diagonal PV correlation for each imaging session. Shaded areas represent mean ± s.e.m. across

sessions. Grey shading indicates reward locations. Two-sided Wilcoxon rank-sum test.

c. Representative CA1 place cells recorded from control and 5xFAD mice (2–4 months). Left:

Normalised two-dimensional spatial rate maps. Right: Linearised positional trajectories showing

place-cell activity across trials. Black lines denote correct trials; red lines denote error trials; black

dots indicate calcium events from place cells.

d. Cumulative distributions with inset box plots showing the median spatial information of CA1

place cells in control and 5xFAD mice.

e–f. Trial-to-trial tuning-curve correlations of place cells. Pearson correlation coefficients plotted

along the linearised trajectory. Shaded areas denote mean ± s.e.m. across cells. Cumulative

distributions of mean diagonal tuning-curve correlations across all place cells with inset box plots

showing the median value for each imaging session.

g. Same analysis as in f, but for in-field versus out-of-field selectivity, calculated as (in-field − out-

of-field) / (in-field + out-of-field).

h. Same analysis as in g, but for the peak in-field activity of place cells.

Statistical comparisons in panels **d–h** were also performed using LMM. Ctrl: n = 4397 recorded

cells, n = 2952 place cells from 20 imaging sessions (10 mice); 5xFAD: n = 3073 recorded cells,

n = 1791 place cells from 13 imaging sessions (7 mice).

3. A stable subset of CA1 place cells is preserved across learning in early-stage

5xFAD mice

The long-term stability of hippocampal place cells is thought to support the spatial

memory retention [35, 36]. To assess whether learning related spatial representation stability

is altered in early-stage 5xFAD mice, we tracked the same CA1 neuronal populations

across three task acquisition stages: early (day 6), middle (day 8), and final (after day 10)

(Fig. 3a). Place cells were identified at each stage and stable place cells were defined as

neurons that (i) exhibited a place field at all stages and (ii) showed tuning-curve (TC)

correlations between adjacent stages exceeding a predefined threshold (>0.51;

Supplementary Fig. 3c; Methods). This approach isolated neurons maintaining consistent

spatial representations across learning (Fig. 3b-d; Supplementary Fig. 3d).

Behavioural alternation performance did not differ between groups (Genotype × Stage: p

= 0.61, LMM; Supplementary Fig. 3a). The proportion of stable place cells relative to all

recorded cells was similar between groups (Ctrl: $14.7 \pm 2.2\%$; 5xFAD: $14.1 \pm 2.9\%$; p =

1; Fig. 3e). Likewise, the proportion of stable cells among all place cells did not differ at

any stage (stage 1: rrb = 0.09, p = 0.89; stage 2: rrb = 0.06, p = 0.89; stage 3: rrb = 0.37,

p = 0.23; Fig. 3e). TC correlations of stable place cells between stages were unchanged

(stage 1-3: rrb = 0.12, p = 0.56; stage 1-2: rrb= 0.007, p = 0.91; stage 2-3: rrb = 0.07, p =

0.25; Fig. 3f). The within session trial-to-trial TC correlations (stage 1: rrb = 0.03, p = 1;

stage 2: rrb = -0.2, p = 0.60; stage 3: rrb = 0.13, p = 1; Fig. 3g), in-field selectivity (stage

1: rrb = -0.02, p = 1; stage 2: rrb = 0.05, p = 0.70; stage 3: rrb = -0.09, p = 0.68), and peak

in-field activity (stage 1: rrb = -0.13, p = 0.46; stage 2: rrb = -0.07, p = 1; stage 3: rrb = -

0.12, p = 0.59; Fig. 3g) were not changed in 5xFAD mice. By contrast, unstable place

cells (Fig. 3d) showed significantly reduced trial-to-trial TC correlations (stage 1: rrb = -

0.06, p = 0.15; stage 2: rrb = -0.18, p < 0.001; stage 3: rrb = -0.13, p < 0.001; Fig. 3h), in-

field selectivity (stage 1: rrb = -0.1, p = 0.04; stage 2: rrb = -0.04, p =0.89; stage 3: rrb =

-0.07, p =0.23) and peak in-field activity (stage 1: rrb = -0.22, p< 0.001; stage 2: rrb = -

0.17, p< 0.001; stage 3: rrb = -0.09, p =0.03) across the majority of learning stages. Thus,

despite spatial coding deficits in unstable place cells, early-stage 5xFAD mice preserve

a subset of stable place cells whose coding properties and prevalence remain intact

across learning.

Figure 3. Spatial representations across training stages in early-stage 5xFAD mice

a. Timeline of imaging sessions during task acquisition, including early-stage imaging (around the

6th training day), middle-stage imaging (around the 8th training day), and late-stage imaging (after

the 10th training day).

b. Example CA1 FOV showing identified place cells at each stage in a 5xFAD mouse. Red

contours indicate stable place cells that maintained stable spatial tuning across all stages (see

methods); grey contours denote unstable place cells; cells without contours were not classified

as place cells. Scale bar: 100 µm.

c. Place fields along the linearised trajectory, sorted by place cell identity at each imaging stage.

d. Same as **c.** but for unstable place cells at each imaging stage.

e. Left: proportion of stable place cells relative to the number of total recorded cells. Right:

proportion of stable place cells relative to the number of total place cells identified at each

individual stage.

f. Tuning curve (TC) correlations for stable place cells between imaging stages (stage1-3; stage1-

2; stage 2-3).

g. Trial-to-trial TC correlations, in-field to out-field selectivity and in-field peak activity for stable

place cells at each imaging stage.

h. Same as **g.** but for unstable place cells.

Data are presented as mean ± s.e.m. Control group: stable place cells, n = 170; unstable place

cells, n = 774, 886, and 945 at each stage; 10 imaging sessions from 5 mice per stage. 5xFAD

group: stable place cells, n = 162; unstable place cells, n = 669, 814, and 732 at each stage; 7

imaging sessions from 4 mice per stage. Statistical comparisons were made using two-sided

Wilcoxon rank-sum tests with Holm–Bonferroni correction for multiple comparisons.

4. Weakened contextual representations in CA1 trajectory-dependent encoding

A splitter cell is a neuron whose activity on a shared path (here, the stem arm of maze) varies depending on the animal's trajectory. Splitter cells are thought to encode task-relevant contextual information and contribute to flexible memory-guided behaviour [37, 38]. To identify splitter cells, we examined neuron with a place field spanning at least five spatial bins along the stem arm and exhibiting trajectory-dependent activity during left and right-turn trials (Fig. 4a; Methods). For each neuron, spatial tuning curves were constructed separately for correct left-right (LR) and right-left (RL) trials. Trajectories were classified as high-rate or low-rate based on which turn direction produced the higher peak spatial activity on the stem arm (Fig. 4b). Splitter cells were then identified based on significant differences between high- and low-rate tuning curves, and splitting quality was quantified using several metrics (Methods).

We found that the spatial tuning calcium activity on the high-rate trajectory was reduced in early-stage 5xFAD mice ($F(1, 14352) = 214.77$, $p < 0.001$; Fig. 4c). The probability that individual stem bin exhibited splitting activity was also decreased in AD mice (Genotype × Spatial bin interaction: $F(15, 14352) = 8.51$, $p < 0.001$; Fig. 4d). Consistently, the proportion of splitter cells was significantly lower in 5xFAD mice compared with controls (Ctrl: 11.9 ± 1.3%; AD: 9.8 ± 2.4%; $p = 0.03$; Fig. 4e).

Next, we evaluated several metrics of splitting quality. Neither split extent ($\beta = -0.05$, 95% CI [-0.12, 0.019], $p = 0.16$; Fig. 4f) nor split discriminability ($\beta = -0.04$, 95% CI [-0.10, -

0.03], p = 0.30; Fig. 4g) differed significantly between genotypes, although both showed

non-significant reduction trends. In contrast, the active split-trial ratio was significantly

reduced in 5xFAD mice (β = -0.12, 95% CI [-0.22, -0.007], p = 0.037; Fig. 4h), the onset

of splitting activity (first split trial) occurred later (β = 1.63, 95% CI [0.95, 2.32], p < 0.001;

Fig. 4i), and trial-to-trial reliability across high-rate trajectories decreased (β = -0.02, 95%

CI [-0.04, -0.008], p = 0.004; Fig. 4j). These results suggest that although the basic spatial

tuning of splitting activity may remain, its trial-by-trial consistency and engagement are

impaired in 5xFAD mice. At the population level, a linear discriminant analysis decoder

trained on all stem-active neurons achieved significantly lower decoding accuracy in AD

mice (correlation rrb = -0.49, p = 0.021; Fig. 4k), demonstrating impaired trajectory-

dependent coding across the network.

Finally, cross-age correlations between splitter cell quality and behavioural performance

further support functional relevance of these neurons (active split-trial ratio: r = 0.50, p =

0.008; split extent: r = 0.57, p = 0.002; split discriminability: r = 0.54, p = 0.004;

Supplementary Fig. 4). Importantly, in early-stage 5xFAD mice, impairments in splitter cell

properties emerged before any observable deficits in task performance in the head-fixed

alternation task.

Figure 4: Impaired hippocampal splitting activity in early stage of 5xFAD mice

a. Two representative stem place cells exhibiting splitting activity in control and 5xFAD mice. From

top to bottom: normalised two-dimensional calcium activity map for the figure-8 maze from one

imaging session; calcium activity raster along the stem for all trials, with the corresponding

linearised behavioural profile; calcium activity raster for left–right trials (blue); calcium activity

raster for right–left trials (red); normalised calcium tuning curves for each turn direction.

b. Difference in calcium activity rates between left and right trajectories for CA1 cells with stem

place fields. Trajectories were classified as high- or low-rate based on the direction with the higher

peak spatial activity rate. Activity was normalised to each cell's peak firing rate and sorted by field

position on the high-rate trajectory. Each line represents one cell.

c. Mean ± s.e.m. spatial activity (from panel b) along the stem for high-rate (solid) and low-rate

(dotted) trajectories.

d. Probability of each stem bin exhibiting splitting activity.

e. The proportion of splitter cells. one-sided Wilcoxon rank-sum test.

f–j. Splitting properties of splitter cells:

f. Proportion of spatial bins showing splitting activity;

g. Discriminability between high- and low-rate trajectories;

h. Proportion of trials with activity on the high-rate trajectory;

i. First trial number exhibiting splitting activity;

j. Trial-to-trial correlation across adjacent high-rate trajectories.

Cumulative plots showed data across all cells, and insert box plot showed the median value of

imaging sessions in control and 5xFAD mice. All assessed with linear mixed-effects models.

k. Decoding accuracy of a discriminant analysis classifier using all stem-active cells. (i) Decoding

accuracy along the stem; (ii) median decoding accuracy for each imaging session. Two-sided

Wilcoxon rank-sum test.

Control: stem place cells, n = 578; splitter cells, n = 521; 20 imaging sessions from 10 mice.

5xFAD: stem place cells, n = 321; splitter cells, n = 264; 13 imaging sessions from 7 mice.

5. Reorganisation of CA1 contextual encoding patterns in early-stage AD mice

Given the pronounced impairments in splitter cell activity in early-stage AD-model mice,

we next examined whether behavioural choice influenced trajectory-dependent encoding

along the stem arm. We focused on CA1 place cells with fields located on the stem and

applied an dissimilarity analysis adapted from Varga et al.[39] to determine whether single-

trial activity reflected the behavioural context of the past, current, or upcoming turn.

Error trials are essential for dissociating retrospective from prospective coding [40],

analyses were restricted which mice had at least three left error and right error trials. For

each spatial bin, we computed dissimilarity of cell activity between two relevant trial types

(e.g., LR vs. LL for retrospective; LR vs. RR for prospective; Fig. 5a) and derived a Retro–

Pro bias index for each cell (Fig. 5b). Consistent with previous studies [39, 40], most splitter

cells in our task exhibited a strong retrospective bias (Fig. 5c).

Despite intact alternation performance at this age, early-stage 5xFAD mice showed

markedly altered distributions of contextual encoding types. The Retro-Pro bias was

significantly shifted toward more prospective-biased encoding compared with controls (β

= 0.13, 95% CI [0.09, 0.18], p < 0.001; Fig. 5d). Retrospective encoding was diminished,

while prospective encoding was enhanced in early-stage AD mice, leading to this

increased Retro-Pro bias during trajectory-dependent contextual coding (Supplementary

Fig. 5). Furthermore, control mice gradually transitioned toward prospective encoding

only when approaching the choice point, but AD mice displayed an earlier and stronger

shift to prospective encoding along the stem (Genotype × spatial bin interaction: $F(15,$

$8779) = 2.29$, $p = 0.003$; two-way ANOVA; Fig. 5e). This pattern reveals a marked

reorganisation of trajectory-dependent contextual coding in early-stage AD mice,

reflected in a shift from retrospective towards more prospective encoding.

Figure 5. Altered hippocampal trajectory encoding patterns in early-stage AD mice.

a. The 1 − (normalized dissimilarity) values for retrospective and prospective encoding are shown

for three example stem arm place cells. Dissimilarity was computed as −log10(p value(Wilcoxon rank-

sum test)) of cell activity at each spatial bin between two trial types (e.g., LR vs. LL for retrospective

encoding; LR vs. RR for prospective encoding); higher dissimilarity indicates greater differences

between trial types.

b. Retro–Pro bias index along stem arm of the cell in panel **a**, ranging from −1 (purely

retrospective encoding) to 1 (purely prospective encoding).

c. Histogram distribution of Retro–Pro bias values in control and AD mice. Most cells in control

mice exhibited a retrospective bias.

d. Average Retro–Pro bias across stem spatial bins for each cell.

e. Retro–Pro bias along the stem arm (16–0 cm toward the choice point). Shaded regions indicate

mean ± s.e.m. across cells. Genotype × spatial bin interaction: $p = 0.003$, two-way ANOVA test;

Ctrl: $n = 384$ splitter cells from 10 imaging sessions (5 mice). AD: $n = 225$ splitter cells from 7

imaging sessions (4 mice).

6. Basal forebrain cholinergic activation improved CA1 coding in early-stage

5xFAD mice

Previous studies have shown that medial septal cholinergic neurons play a key role in

enhancing hippocampal theta oscillations during exploration[41]. We therefore examined

whether activating these neurons could enhance spatial and contextual coding in

hippocampal CA1 and mitigate early deficits in 5xFAD mice. To achieve selective

activation, AAV-hSyn-DIO-hM3D(Gq)-mCherry was injected into the medial septum of

ChAT-Cre[+]/5xFAD[+] (AD) and ChAT-Cre[+]/5xFAD[-] (control) mice, and AAV-syn-

jGCaMP7s was injected in dorsal CA1. A cranial window was then implanted for in vivo

calcium imaging (Fig. 6a). After all mice reached the task acquisition criterion, vehicle

(saline) was administered intraperitoneally on the first imaging day and CNO on the

second imaging day. CNO administration did not alter behavioural alternation

performance and locomotion in either group (Ctrl: p = 0.81; AD: p = 0.64; two-sided

Wilcoxon signed-rank test; Fig. 6b, Supplementary Fig. 6a,b).

Tracking the same neuronal populations across vehicle and CNO sessions, we observed

increased global spatial information (computed from two-dimensional spatial firing maps)

following CNO administration in both control (β = 0.37, 95% CI [0.13, 0.61], p = 0.002)

and 5xFAD mice (β = 0.30, 95% CI [0.06, 0.54], p = 0.013). Similarly, place cell spatial

information during correct trials increased in both groups (Ctrl: β = 0.12, 95% CI [0.008,

0.23], p = 0.036; AD: β = 0.19, 95% CI [0.06, 0.32], p = 0.005). In 5xFAD mice, this was

accompanied by a significant increase in trial-to-trial TC correlation of place cells following

CNO administration (β = 0.02, 95% CI [0.004, 0.036], p = 0.015; Fig. 6e,f), whereas

control mice showed no such change (β = 0.00, 95% CI [-0.01, 0.01], p = 0.95). However,

CNO treatment did not rescue place cell in-field selectivity or peak in-field activity in

5xFAD mice (Supplementary Fig. 7b-d), although we observed a reduction in place field

size in control mice after CNO (Supplementary Fig. 6e).

We next assessed trajectory-dependent activity in stem place cells. Mirroring the

improvements observed in place cells, the ratio of trials in which stem place cells

displayed trajectory-dependent coding increased significantly in 5xFAD mice after CNO

administration (β = 0.15, 95% CI [0.05, 0.26], p = 0.003; Fig. 6g). Moreover, the first-trial

onset of splitting activity exhibited a trend towards earlier emergence in 5xFAD mice (β =

-1.25, 95% CI [-2.50, 0], p = 0.050; Fig. 6h), whereas no significant changes were

observed in control mice (splitting extent: β = -0.04, 95% CI [-0.11, 0.036], p = 0.32; first

split trial: β = -0.02, 95% CI [-1.46, 1.40], p = 0.97). Other metrics of trajectory-dependent

activity, including extent, reliability, and discriminability, were unaffected by CNO in

5xFAD mice (Supplementary Fig. 6g-i).

Together, these findings suggest that chemogenetic activation of medial septal

cholinergic neurons enhances the spatial representations and promotes trajectory-

dependent coding in CA1, thereby increasing trial-wise stability in early-stage 5xFAD

mice (Supplementary Fig. 6j), although the rescue remains partial and does not extend

to all dimensions of hippocampal coding.

Figure 6. Basal forebrain cholinergic activation modulated CA1 spatial and contextual coding in early-stage 5xFAD mice

a. Experimental schematic showing DREADD (hM3Dq) expression in medial septal cholinergic neurons of the basal forebrain and jGCaMP7s expression in hippocampal CA1. Right: representative image of hM3Dq expression in ChAT-Cre mice. Scale bar, 100 μm.

b. Performance in a spontaneous alternation task following intraperitoneal (i.p.) injection of saline (vehicle) or clozapine-N-oxide (CNO); Wilcoxon signed-rank test.

c. Global spatial information (from two-dimensional spatial firing maps) of all CA1 neurons after vehicle or CNO administration.

d. Spatial information of CA1 place cells during correct trials following vehicle or CNO treatment.

e. Trial-to-trial tuning curve correlations of CA1 place cells during correct trials under vehicle or

CNO conditions.

f. Mean diagonal Pearson correlation coefficients of CA1 place cells from (**e**).

g. The proportion of trials in which stem place cells displayed trajectory-dependent coding.

h. Latency to the first trial exhibiting trajectory-dependent activity in stem place cells. For clarity,

outliers >10 are not shown in figure.

Data were analysed using linear mixed-effects models with treatment group (CNO vs vehicle) as

a fixed effect and Mouse ID as a random intercept. Ctrl: 964 detected neurons; place cells: 697

(140 stem place cells) after vehicle, and 522 (122 stem place cells) after CNO; 5 mice. AD: 777

detected neurons; place cells: 522 (46 stem place cells) after vehicle, and 449 (41 stem place

cells) after CNO; 4 mice.

Discussion

In this study, we introduced a novel head-fixed behavioural paradigm to assess hippocampal-associated spatial memory. This paradigm allowed us to examine behavioural performance and hippocampal neuronal spatial representations by two-photon in vivo imaging in 5xFAD mice during a continuous spatial alternation task (Fig. 1a, b). Previous research has established that the 5xFAD mouse begins to develop amyloid deposits between 2 to 4 months of age, with intraneuronal accumulation of Aβ42 starting as early as 1.5 months [29] . This amyloid burden in 5xFAD mice intensifies over their lifespan [29, 42]. By 4 to 6 months, these mice exhibit substantial amyloid deposition alongside observed deficits in synaptic transmission, LTP, and sharp wave-ripple events [26, 29, 33, 43]. While some studies have detected mild impairments in spatial or hippocampus independent memory as early as 3 to 4 months, more pronounced memory deficits become evident at later stages, beyond 4-6 months of age [27, 44, 45, 46]. Consistent with these findings, our study observed intact behavioural performance in young 5xFAD mice (2-4 months) during the head-fixed alternation task (Fig. 1c-e), and substantial behavioural impairments emerged only after ~7 months of age in 5xFAD mice (Fig. 1f–h).

Disruptions in hippocampal spatial representations have been reported in both human sporadic AD and multiple AD mouse models [47, 48]. In our study, although 2-4 month-old 5xFAD mice exhibited normal performance in the spatial alternation task (Fig. 1c–e), their CA1 neurons showed clear impairments in spatial coding, including reduced spatial

information and decreased stability at both the population level and in task-related place

cells (Fig. 2). Interestingly, global spatial representations were preserved, yet task-related

spatial coding was impaired, suggesting that behavioural demands reveal spatial

information deficits that are more vulnerable at the global level. Notably, a previous report

using 3-5 month-old APP-KI mice found intact CA1 spatial tuning and remapping despite

the presence of amyloid plaques [49]. These differences may reflect model-specific

progression of pathology or the fact that our task is strongly goal-directed, which places

greater demands on task relevant place cell coding.

The stabilisation of place cell coding is closely linked to the memory demands and

attentional engagement required in goal-directed tasks [50, 51]. In our task, stable place cells

in early-stage 5xFAD mice continued to exhibit consistent spatial tuning across learning

stages (Fig. 3c, e-g). Previous work has shown that more excitable neurons tend to

maintain stable spatial representations across days [52], and that stable place cells

accumulate during learning correlate strongly with behavioural performance in healthy

mice [53]. Thus, the preserved quality of stable place cells in young 5xFAD mice may help

maintain long-term spatial memory across repeated training in a familiar environment,

thereby supporting intact performance in the spatial alternation task. In contrast, CA1

population activity showed reduced flexibility: PV correlations across learning stages were

higher in 5xFAD mice (Supplementary Fig. 3b), indicating diminished population-level

remapping and a more rigid representational structure across task learning. Together,

these results suggest that although broader CA1 coding is destabilised in early-stage AD,

a preserved subset of stable place cells may provide sufficient representational continuity

to sustain behaviour.

Recent work suggests that the hippocampus supports cognitive processes beyond spatial

mapping [54, 55, 56]. Our findings extend this view by demonstrating impairments in both

spatial and contextual representations within the same behavioural framework. Splitter

cells, which encode past choices and future behavioural outcomes, are thought to

integrate information within prefrontal-thalamo-hippocampal circuits that guide flexible

navigation [38]. Trajectory-dependent coding was markedly degraded, at populational level

and single cell level at early-stage of AD (Fig. 4). Although the splitting extent and

discriminability were preserved in 5xFAD mice, the trial-trial engagement and stability of

splitting activity were reduced and closely associated with behavioural deficits (Fig. 4e-

j;Supplementary Fig. 4). Notably, while the proportion of place cells showed a mild, non-

significant reduction, the proportion of splitter cells was significantly diminished, indicating

that trajectory-dependent representations are more vulnerable than spatial encoding at

early stages of AD.

We found that most splitter cells preferentially encoded retrospective rather than

prospective information in our task (Fig. 5c,d), indicating that neuronal activity along the

common stem was dominated by information about the past choice. This retrospective-

prospective balance became more even only when mice approached the choice point

(Fig. 5c,d). The earlier and stronger shift toward prospective coding observed in 5xFAD

mice may reflect weakened maintenance of past information before the next decision,

potentially linked to reduced hippocampal-prefrontal synchrony reported in 5xFAD mice

[57]. Previous studies have similarly shown that splitter cells typically favour retrospective

encoding, and our results are consistent with this pattern. However, the stronger

retrospective bias near the choice point in our task may be influenced by the shorter stem

length compared with earlier studies [39, 40].

A reduction in trial-to-trial stability was evident in both spatial and contextual

representations, suggesting that early coding deficits in CA1 may reflect a common

disturbance affecting representational consistency. These spatial and contextual coding

impairments may link with known synaptic disturbances in 5xFAD mice, including reduced

long-term potentiation (LTP) and impaired synaptic transmission from early age [17, 26, 43,

58]. Changes in CA1 representations across learning rely heavily on synaptic plasticity

driven by inputs from entorhinal cortex layer 3 [59]. The emergence of splitter cells likely

requires the recruitment of more complex and highly integrated neural networks, as well

as stronger plasticity mechanisms. Given that synaptic plasticity can convert previously

silent CA1 neurons into splitter cells [23], early synaptic impairments may therefore

contribute to deficits in both spatial and contextual representations observed in 5xFAD

mice.

Chemogenetic activation of medial septal cholinergic neurons enhanced spatial

information and increased the trial-wise engagement of spatial and contextual encoding

in early-stage 5xFAD mice (Fig. 6). This system is known to be compromised in AD, and

cholinergic decline has been linked to cognitive deterioration in several

neurodegenerative conditions[60, 61, 62]. Basal forebrain cholinergic neurons send dense

projections to the hippocampus and play a central role in regulating hippocampal synaptic

plasticity, learning and memory[28]. Cholinergic activity also promotes hippocampal theta

oscillations, which support spatial coding and memory-guided behaviour [63]. In 5xFAD

mice, the basal forebrain is the first brain region to show cholinergic neuron loss,

beginning to degenerate from around nine months of age, which indicates an early

vulnerability of this system [64]. Although Alzheimer's disease remains incurable, several

approved treatments, including donepezil, galantamine and rivastigmine, act as

cholinesterase inhibitors and increase acetylcholine levels. These observations suggest

that targeting basal forebrain cholinergic neurons may still offer a promising strategy for

early intervention in AD.

Early tetrode studies were limited by the small number of neurons that could be recorded

simultaneously, whereas two-photon imaging in virtual-reality environments, although

scalable, lacks the multimodal sensory cues that shape place-cell activity [65, 66]. More

recent real-world head-fixed mazes address some of these constraints [67] , yet most

imaging studies still use relatively simple running tasks. Because such tasks place limited

cognitive demands on the animal, they may not fully reveal the types of representational

deficits relevant to AD. This study integrates a real-world environment with a cognitively

demanding alternation task to investigate hippocampal function in early-stage AD,

providing insights under behaviourally relevant context.

Limitations of the study

First, we did not account for sex differences in our analysis. Human studies indicate that

AD prevalence and severity are higher in females, with sex-specific differences observed

in amyloid pathology and cognitive decline [68, 69]. Similar patterns have been reported in

the 5xFAD model, with female mice exhibiting greater amyloid deposition and potentially

distinct cognitive impairments compared to males [70, 71, 72, 73]. Second, although calcium

imaging is a powerful tool for assessing neuronal activity, potential discrepancies in

calcium indicator dynamics in AD models must be considered. Variations in calcium

indicator expression, deregulation of calcium signalling, and altered decay kinetics in

5xFAD mice could affect the accuracy of spike inference [74]. Third, incorporating pupil and

lick tracking could provide further insights into attention and task engagement, thereby

clarifying the mechanisms underlying the observed behavioural impairment. Finally, our

behavioural assessment was limited to a single task that lacked an explicit delay period,

relying instead on the animal's natural return time to the choice point. Incorporating a

controlled delay (such as with a servomotor gate) could increase the task's reliance on

intrinsic hippocampal processing, as suggested by studies in freely moving animals, but

was beyond the scope of the present work [37].

Materials and Methods

Experimental Animals:

5xFAD mice express human APP and PSEN1 transgenes with a total of five AD-linked mutations: the Swedish (K670N/M671L), Florida (I716V), and London (V717I) mutations in APP, and the M146L and L286V mutations in PSEN1. Male hemizygous 5xFAD mice (B6SJL-Tg (APPSwFlLon, PSEN1*M146L*L286V) 6799Vas/Mmjax, JAX stock # 034840) were crossed with female homozygous Chat-IRES-Cre($\triangle$neo) mice (JAX stock # 031661 on a C57BL/6J genetic background) to maintain the 5xFAD positive (5xFAD$^+$/Chat-Cre$^+$ as 5xFAD group) and non-5xFAD (5xFAD$^-$/Chat-Cre$^+$ as control group) mice. Colonies under standard animal breeding at Imperial College London. The animal room was kept on a reverse 12:12 light-dark cycle daily.

Ear samples were collected from 5xFAD-Tg and Non-5xFAD-Tg transgenic littermates for genotyping identification. Real-time PCR was used for genotyping 5xFAD transgenic with probes APPswTg and huPSEN1Tg, conducted by Transnetyx. Animals were divided into two age groups: young group aged (young control median age = 3.6 months, N=10; young AD median age = 3.6 months, N=7) and old group aged (old control median age = 8.6 months, N=6, old AD median age = 7.9 months, N=6). Non-AD transgenic littermates were used as controls. Both male and female mice were used for this study. All experimental procedures were carried out under the Animals (Scientific Procedures) Act 1986 and according to Home Office and institutional guidelines, under a Project License to SR

Schultz.

Surgeries:

Mice were anaesthetised with 1.5-3% isoflurane (Iso-Vet, Chanelle Pharma) and body

temperature was maintained at 37°C. Analgesia was administered pre-operatively with

Carprofen (5 mg/kg SC, Rimadyl, Zoetis) and Buprenorphine (0.07 mg/kg SC, Vetergesic,

Ceva Animal Health Ltd). Anaesthetic depth was monitored every 10 minutes via the

pedal flexion reflex. A small (0.5 mm) craniotomy was performed and 1 μl of a viral cocktail

containing two adeno-associated viruses (pGP-AAV-syn-jGCaMP7s-WPRE, Addgene

catalogue 104487-AAV9; pAAV-CAG-tdTomato, Addgene catalogue 59462-AAV9) in a

3:1 ratio was injected into the hippocampal CA1 region (coordinates from bregma, in mm:

-1.3 and -1.5 DV, -1.8 ML, -2.0 AP). Ten minutes after viral injection, a 3-mm circular

craniotomy was made, and the overlying cortex (including parietal cortex and parts of

visual and hindlimb sensory cortex) was aspirated using a 27-gauge needle connected to

a water pump until the corpus callosum fibres were exposed. A stainless-steel cannula (3

mm diameter, 1.5mm height) with a glass bottom was inserted and secured with histoacryl

glue (B. Braun Surgical). The hippocampal imaging window was described by Dombeck

et al. [75]. A stainless-steel head plate (8.5 mm aperture) was glued to the skull, centred

over the craniotomy. The exposed skull outside the head plate aperture was covered with

dental cement mixed with black powder paint. Postoperatively, mice received Carprofen

(5 mg/kg/24 hrs) in oral water for three days. Behavioural training began after a 2-3 week

recovery period.

Behavioural training in head-fixed mice

Maze apparatus:

We used a carbon fibre maze with attached visual and tactile cues, floating on the

Neurotar tracker air table (Mobile HomeCage, Neurotar). Built-in magnets in maze

allowed for tracking the animal's position within the environment. The maze has a

diameter of 32.5 cm, with a 16 cm-long stem arm. The corridor width for all arms was 5

cm. Visual cues were provided by attached phosphorescent tape, and tactile cues were

added using Fiberglass Mesh Tape on the inner wall of the maze. The training cage tape

was fully charged with daylight before the commencement of any training sessions.

Handling and habituation:

Acclimatization to both the room and maze is essential for mice to become familiar with

the training environment prior to official training sessions. Mice were handled at least two

days before behavioural training. On the day before training began, mice freely explored

the air-lifted T maze in the training room under red lighting for 10 minutes. During the first

5 training days, mice were allowed 5 minutes of free exploration before head fixation.

Spatial continuous rewarded alternation task:

To motivate task engagement, mice were subjected to water restriction prior to the first

day of training. After 5 days, their body weight typically stabilised at 85 to 90% of initial

body weight, with each mouse receiving a minimum of 1.5 ml of water per day. If the full

daily water ration was not received during the task training, supplementary water was

provided in the home cage after the second session to maintain body weight above 85%.

Mice were trained in the dark during the learning of the alternation task. For the first trial,

mice could freely choose reward arm. A 4 µl reward was delivered via a lick spout

attached to the head-plate clamp and controlled by a peristaltic pump (Campden

Instruments), accompanied by a beep. In subsequent trials, a reward was only delivered

when mouse entered the reward arm opposite to the one visited in the previous trial. A

complete trial required the mouse to pass through the previous trial reward position, return

arm, stem arm, choice arm, and arrive at the next reward location. Head-fixed animals

were trained for 45 minutes each session twice a day, with at least 4 hours between

sessions. From the 6th training day onward, 1% sucrose water was used as the reward

during training and imaging sessions, but still supplied water for supplementary fluid.

Mouse position and behavioural data were recorded using the Neurotar Mobile

HomeCage tracker (V3.0.0.64). The figure-8 maze was cleaned with 70% ethanol before

each mouse training or imaging recording.

In vivo two-photon Ca2+ imaging:

Two-photon calcium imaging was conducted using a Scientifica resonant scanning

microscope with a 16× Nikon water-immersion objective (LWD 0.8 NA). jGCamp7s and

tdTomato fluorophores were excited at 940 nm using Ti:Sapphire laser (Mai Tai HP,

Newport). The 50% (vol%) ultrasound gel was used as the immersion medium. Images

(490 × 490 µm, 512 × 512 pixels) were acquired at 30.9 Hz frame rate using SciScan 1.3

software. Imaging was synchronized with behaviour recording via a start TTL signal sent

from the behaviour recording software Neurotar. All mice underwent imaging afterwards

10th training day. Additionally， some mice were also imaged on 6th and 8th training

days to enable imaging across the learning period. Each imaging session lasted no longer

than two hours per mouse per day. At the beginning of each session, mice were head-

fixed at the stem arm and restricted by two carbon fibre doors. Once the targeted field of

view (FOV) was located, the doors were removed and imaging commenced as the mouse

began running. To ensure sufficient data for spatial tuning analyses, a minimum of 80

total trials was required for each imaging session, though approximately 100 trials were

usually recorded per session.

Chemogenetic activation

For chemogenetic activation of basal forebrain cholinergic neurons, 0.5–1.0 µL AAV-

hSyn-DIO-hM3D(Gq)-mCherry was injected into the medial septum (AP +0.8 mm, ML

+0.1 mm, DV −4.2 to −4.0 mm from bregma), and jGCaMP7s was delivered into dorsal

hippocampal CA1. DREADDs experiments were performed only after mice reached

stable task performance. Animals were recorded across two consecutive imaging days.

On the first day, mice received an intraperitoneal injection of saline (vehicle). On the

second day, Clozapine-N-oxide (CNO; 1 mg/kg, i.p.; water-soluble CNO dihydrochloride,

HelloBio HB6149) was administered 40 minutes before imaging. A low CNO dose was

used to minimise off-target effects, and previous studies have reported that 1 mg/kg CNO

does not alter hippocampal neuronal activity without hM3D injection in ChAT-Cre mice [76].

The same neuronal population was tracked across vehicle and CNO sessions.

Quantification and statistical analysis

Behaviour data analysis:

Only complete trials (see behaviour training) were included in the analysis. As mice

received a reward on their initial free choice, this first choice was excluded from the

calculation of correct trials. Alternation performance was defined as the percentage of

correct trials out of the total number of trials. To examine changes in performance across

training days, we fitted a linear mixed-effects model (LMM): data ~ 1 + Genotype × Day

+ (1 | MouseID). Performance (percentage of correct trials) was the dependent variable.

The model included Genotype (5xFAD, Control), Training days (as a continuous

numerical predictor), and their interaction as fixed effects. To account for repeated

measures from the same subject, Mouse ID was included as a random intercept. To

further assess learning dynamics independent of daily variability, trials from all sessions

were concatenated and binned into consecutive 50-trial blocks, with the percentage of

correct responses computed for each block. Learning trajectories across groups were

then modelled using a quadratic mixed-effects model: Performance ~ Genotype × (Block

+ Block²) + (1|MouseID). Differences in task learning performance between 5xFAD mice

and their age-matched controls were determined based on the Genotype × Day and

Genotype × Block interaction effects. Task acquisition criterion is the percentage of

correct trials is over 70% over four consecutive trial blcoks.

Trajectory linearisation:

To linearise the mouse trajectory, the maze was segmented into four functional zones:

the return arm, stem arm, choice arm and reward arm. Trajectories were parsed

accordingly, starting from the return arm, then moving up the stem arm into the choice

arm, and ending in the reward arm. The reward spout marked the boundary between the

choice and reward arms. Each trajectory was then categorised into L-R, R-L, L-L and R-

R trial types based on maze segmentation. Linearised trajectories were mapped by trial

type onto a standardised 49.5 cm track, corresponding to the estimated length of a

complete trial. Trials were chronologically ordered based on the sequence of completion

in the maze.

Imaging data processing:

To extract calcium activity from individual neurons, we employed a customised image

processing pipeline based on the CaImAn MATLAB package [77], following the protocol

described by Go et al. [67]. Imaging data were initially corrected for motion artefacts using

sequential rigid and non-rigid registration methods, aligning frames a reference template

from the same imaging session. Regions of interest (ROIs) were then identified, with

excluding any overlapping ROIs. To enable tracking of ROIs across multiple imaging

stages, images from different stages were registered to a template from one session, and

ROIs were identified on a temporally concatenated video. For quantification of single-

neuron calcium amplitude trains, neural activity was deconvolved using a sparse non-

negative deconvolution approach. This was implemented with an exponential kernel

corresponding to the decay timescale of the calcium indicator employed, utilizing the

OASIS (Online Active Set method to Infer Spikes) algorithm to infer fluorescence activity

into calcium trains with both onset times and amplitudes.

To eliminate out any discrepancies in OASIS event amplitude between different mice and

FOVs, we fitted the distribution of event amplitudes for each FOV in each mouse to a

lognormal distribution and standardised the distribution for each FOV to unit mean. This

eliminates variability in event amplitude due to differences in signal to noise ratio (for

instance due to imaging depth and viral expression), and makes it possible to compare

calcium activity rates objectively across FOVs and animals. The resulting deconvolved

calcium event amplitudes are expressed in consistent arbitrary units across the entire

dataset.

Spatial information:

To quantify spatial representations in CA1 neurons, spatial information was calculated

from two-dimensional spatial rate maps for each detected neuron. The figure-8 maze was

discretised into 1 cm-wide bins, forming a 32 × 32 spatial grid. For each bin, the spatial

activity rate was computed as the sum of standardised deconvolved calcium amplitudes

recorded within that bin, divided by the total occupancy time, considering only periods

when the mouse's running speed exceeded 8 mm/s. The resulting spatial rate maps were

smoothed using a Gaussian kernel with a sigma of [3, 3]. The spatial information content

of each neuron was calculated using Skaggs's spatial information method [78] (bits per arb,

standardised amplitude rate), is calculated using the following formula:

$$SI = \sum_{i=1}^{N} P_i \frac{\lambda_i}{\lambda} \log_2 \left(\frac{\lambda_i}{\lambda} \right)$$

where SI is the spatial information in bits/arb. P_i is the probability of the mouse being in

the i-th bin. λ_i is the mean standardised amplitude rate of the cell when the mouse in the

i-th bin. λ is the overall mean standardised amplitude rate of the cell across the entire

environment. N is the total number of spatial bins.

Population vector correlation:

To assess the stability of spatial representations at populational level, we calculated

population vector (PV) correlations between odd and even trials. For each session, we

first constructed spatially binned activity curves for all detected cells. The population

vector at each spatial bin was defined as a vector containing the calcium activity rates of

all cells within that bin. We then computed the PV correlation matrix by comparing

population vectors between odd and even trials across all spatial bins by Pearson

correlation. The diagonal elements of this matrix represent the correlation coefficients for

the same spatial positions between trial types, reflecting the stability of spatial coding

along the linearized trajectory. Finally, we averaged the diagonal PV correlation

coefficients across animals within each experimental group.

Place cell identification:

Place cells from two-dimension spatial maps were identified based on the following

criteria: 1) A neuron's SI must exceed the 95th percentile of the shuffled SI values. 2) the

neuron should be active within the place field for more than 2% of the time the mouse

spends in that field. 3) the size of the place field must exceed 5 cm^2. The boundary of a

place field was determined by encompassing all connected occupancy bins where was

at least 50% of the peak rate in the occupancy bin.

For place cells selection from one-dimension spatial maps, we calculated for correct trials

(R-L and L-R) and error trials (L-L and R-R) based on same type of linearised trajactory.

The criterions for 1D place cells : 1) A neuron's SI must exceed the 95th percentile of the

shuffled SI values. 2) the neuron should be active within the place field for more than 1%

of the time the mouse spends in that field. 3) the neuron were active more than 30% of

total number of trials.

Place cell quality metrics:

In-field selectivity: the in-field and out-of-field activity rates were computed and

normalized by the number of spatial bins. Selectivity was defined as (In-field – Out-

field)/(In-field + Out-field). ranging from −1 (purely out-of-field firing) to 1 (purely in-field

firing). Peak in-field activity: the maximum activity value within each place field,

determined by the spatial bin with the highest activity rate. We also computed: Active trial

ratio (Supplementary figure 2): the proportion of trials of the same trajectory type (i.e. left-

to-right, or right-to-left) in which a given place cell was active. Adjacent trial-to-trial

reliability: for each trial, the spatial activity profile of the neuron was constructed by binning

calcium activity along linearised trajectory. Pairwise Pearson correlations were then

computed between adjacent active trials of the same trajectory type. The trial-to-trial

reliability of each neuron was defined as the mean of these correlation coefficients across

all active trial pairs.(for place cells have multiple place field, the PF selectivity and infield

peak was computed on the average of the multiple place field)

Trial-to-Trial Tuning curve (TC) correlation:

For each identified place cell, we computed trial-to-trial tuning curve correlations to

quantify response reliability. Single-trial tuning curves were constructed based on

spatially binned calcium activity. Trials were divided into odd and even subsets, and

Pearson correlations were calculated between corresponding spatial bins across these

trial groups. The resulting correlation matrix's diagonal elements, representing reliability

at same spatial positions, were averaged to obtain a single stability metric per cell.

Tuning curve (TC) correlation across stages:

TC correlations were computed only for neurons classified as place cells. For each

imaging stage, spatially binned activity along the linearised trajectory was constructed for

every place cell. TC correlations between stages were calculated using Pearson's

correlation coefficient. For stage-1 correlations, tuning curves were generated separately

from odd and even trials within the same stage, and the correlation between these two

curves was used as the within-stage TC correlation.

Identification of stable place cell across stages:

Stable place cells were defined as neurons that (i) exhibited a place field at all imaging

stages and (ii) showed tuning-curve (TC) correlations between adjacent stages exceeding

a predefined threshold (0.51). The stability threshold (0.51) corresponded to the 95th

percentile of TC correlations obtained from 1,000 randomly paired place cells at stage

1(Supplementary Fig. 3c).

Splitter cell identification:

To identify hippocampal splitter cells, the stem arm of the figure-8 maze was divided into

1-cm spatial bins. Place cells with place fields spanning at least five consecutive bins

along the stem were included for analysis. For each cell, neural activity was computed at

each stem bin across trials, and spatial tuning curves were constructed separately for

correct left-turn and right-turn trials. The difference between the two tuning curves was

then calculated at each bin. Significance was assessed using a permutation procedure in

which left- and right-turn trial labels were randomly shuffled 1,000 times. A cell was

classified as a splitter cell if (i) it possessed a place field covering at least five stem-arm

bins and (ii) its real tuning-curve difference exceeded the 95th percentile of the shuffled

distribution in at least one spatial bin.

To ensure that splitting activity did not arise from stereotyped speed differences along the

stem, we evaluated whether each spatial bin exhibited speed-independent splitting. For

each neuron, we first constructed spatially binned activity for every trial. We then

performed a linear mixed-effects analysis (MATLAB fitlme) for each stem place cell

(defined as cells with a place field > 5 cm along the stem). The model included trial type

(left/right) as a categorical predictor and animal speed as a continuous predictor, with

spatially binned activity as the dependent variable. A spatial bin was considered to exhibit

valid splitting activity if trial type remained a significant predictor of neuronal activity after

controlling for speed. Neurons with at least one spatial bin meeting this criterion were

classified as splitter cells.

Splitter cell quality metrics:

The quality of neuronal splitting activity was assessed using five metrics, computed for

identified splitter cells:

1) Split extent [79]

The proportion of spatial bins along the stem arm that exhibited significant splitting activity.

2) Split discriminability.

Split discriminability was calculated as the sum of the absolute differences between left-

and right-turn tuning curves across all stem bins, normalised by the total summed activity

of both tuning curves across those bins.

3) Split reliability (Trial-to-trial correlation)

This metric quantified the consistency of splitting activity across trials. For each cell,

Pearson correlation coefficients were computed between the spatial activity profiles of

adjacent trials in the neuron's preferred trajectory direction (e.g., between consecutive

left–right trials if the cell preferred the L-R trajectories). Correlations were calculated only

within the same imaging session.

4) Active split-trial ratio.

The proportion of trials in which a splitter cell was active during its high-rate trajectory

direction.

5) First split trial

The trial number at which a cell first is active on its splitting spatial bin.

Discriminant classification:

A linear discriminant classifier was trained using 70% of randomly selected correct trials

from each imaging session. The classifier was implemented using MATLAB's built-in

function fitcdiscr. For each neuron, calcium activity at each time point while the mouse

occupied the stem arm served as input features, and the mouse's upcoming turn direction

was used as the response variable. Only correct trials (left-right and right-left) were

included in both the training and testing sets. The trained classifier was then used to

predict turn direction on the remaining 30% of correct trials. This procedure was repeated

1,000 times, each with different randomly selected training and testing subsets. Decoding

accuracy was computed within 1 cm spatial bins along the stem arm, and session level

decoding accuracy was defined as the mean accuracy averaged across all spatial bins.

Quantification of Retrospective-Prospective Bias:

We quantified the influence of past versus future behavioural choices on neuronal activity

by computing a dissimilarity-based Retrospective-Prospective (RP) bias index, adapted

from previously established approaches Varga et al.[39]. Analysis was restricted to

recording sessions containing at least three error left trials and three error right trials. For

each spatial bin along the stem, we extracted calcium activity rates from four trial types:

RL: correct left; LR: correct right; LL: error left; RR: error right. For each spatial bin,

dissimilarity between two trial types was computed as: $D = -\log10(p)$, where p is the

Wilcoxon rank-sum test p-value comparing firing-rate distributions across trials. Higher

dissimilarity values indicate greater difference in neural activity patterns between the two

compared trial types. Four pairwise comparisons were performed: Retrospective

comparisons (LL vs. LR; RR vs. RL), Prospective comparisons (LL vs. RL; RR vs. LR).

For each spatial bin, we derived: $D_{retro}=\min[D(LL,LR),D(RR,RL)]$; D_{pro}

$=\min[D(LL,RL),D(RR,LR)]$. The Retro–Pro (RP) bias index was then computed as: $(D_{retro}$

$- D_{pro})/ (D_{retro} + D_{pro})$. This value ranges from −1 (purely prospective encoding) to +1

(purely retrospective encoding), with values near zero indicating balanced encoding.

Statistical tests

All statistical analyses and visualisation were performed in MATLAB. Data are presented

either as mean ± s.e.m. or as box plots. The significance threshold was set at $p \leq 0.05$.

Behavioural performance across training days or trial blocks was analysed using linear

mixed-effects models (LMMs). Genotype, Day (or Block) and their interaction (Genotype

× Day/Block) were treated as fixed effects, and Mouse ID was included as a random

intercept to account for repeated measures. For datasets with multiple imaging sessions

per mouse (where each session corresponded to a distinct imaging field of view, with no

repeated cells across sessions), linear mixed-effects models included nested random

intercepts for MouseID and SessionID to account for within-animal and within-session

variability (data ~ 1 + Genotype + (1 | MouseID:SessionID)) . For datasets in which each

mouse contributed only a single imaging session (e.g., DREADD experiments), only

MouseID was included as a random effect, as SessionID and MouseID were equivalent

in this context. Stable place cells across learning stages were pooled across mice and

analysed using non-parametric comparisons. For non-parametric comparisons, effect

sizes were reported as rank-biserial correlation (rrb). For linear mixed-effects models,

estimates are reported with 95% confidence intervals. For all other comparisons, non-

parametric Wilcoxon rank-sum or Wilcoxon signed-rank tests were used to avoid

assumptions of normality, unless otherwise specified in the relevant methods. Data points

are shown whenever possible, and Bonferroni corrections were applied for multiple

comparisons. For datasets displayed as box plots, boxes indicate the interquartile range

(IQR; 25th–75th percentiles), the central line denotes the median, and whiskers extend

to 1.5 × IQR; points beyond this range are plotted as outliers.

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Y.L. and S.R.S. conceived and designed the study. Y.L.mainly conducted experiments

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analyzed data. M.A.G. and H.Z. contributed to imaging data processing. Y.L. and S.R.S.

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Supplementary Information

Supplementary Figure 1. Head-fixed behavioural training in 5xFAD mice (related to Figure 1)

a. Representative *in vivo* imaging of amyloid plaques in hippocampal CA1 regions, labelled with Methoxy-X04. Quantification shows the percentage area of amyloid plaques relative to the total field of view (FOV) in 5xFAD mice aged 2–4 months (*n* = 3 mice) and 7–12 months (*n* = 3 mice). Data are presented as mean ± s.e.m.; comparisons were performed using a two-sided Student's *t*-test. Scale bar, 100 μm.

b. *Left:* Mean total number of trials completed per training day (mean ± s.e.m.) for 2–4-month-old mice, analysed using a linear mixed-effects model (LMM). *Right:* Cumulative number of trials completed by each mouse across all training sessions. Statistical comparisons were performed using a two-sided Wilcoxon rank-sum test.

c. *Left:* average running speeds across all correct trials across training sessions. *Right:* average running speeds across all error trials across training sessions. Group differences were evaluated using LMM.

d. Alternation performance across training days segmented into 50-trial blocks (up to 30 blocks) for 2–4-month-old mice. Individual mouse performance is shown as lines.

e. Proportion of mice reaching the acquisition criterion (>70% correct across four consecutive trial blocks) and the corresponding trial block at which criterion was achieved. Statistical significance was assessed using a log-rank test.

f-i. Same analyses as in panels **b-e** but for mice aged 7–12 months.

Supplementary Figure 2 | Behavioural performance during imaging and CA1 place cell properties in correct versus error trials (related to Fig. 2)

a-e. Behavioural measures per imaging session: (a) alternation accuracy, (b) total trials, (c) correct trials, (d) mean speed including speeds below 8 mm/s, and (e) mean speed during correct trials above 8 mm/s. Two-sided Wilcoxon rank-sum tests.

f. Left: cumulative distribution of active-trial ratio, defined as the proportion of trials in which a cell was active within its place field. Inset shows session medians. Right: active-trial ratio across linearised maze position.

g. Same as f, showing adjacent trial-to-trial tuning curve correlation (pairwise Pearson's r), calculated only from trials where the cell was active. LMM used for Ctrl vs AD comparison.

h-k. Place cell properties in correct and error trials: (h) active-trial ratio, (i) trial-to-trial reliability, (j) in-field versus out-of-field selectivity, and (k) peak in-field activity. Two-sided Wilcoxon signed-rank tests.

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41 **I.** Mean trial speed per session during correct and error trials.
42 Control group includes 2952 cells from 20 sessions (10 mice), 5xFAD group includes 1791 cells
43 from 13 sessions (7 mice).
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Supplementary Figure 3. Place cells across training stages (related to Figure 3)

a. Behavioural alternation performance across imaging stages. The grey dotted line marks the 70% performance criterion. LMM test.
b. PV correlation of all cells between stages.
c. Tuning curve (TC) correlation threshold used to define stable place cells across days. Stage-1 TC correlations were computed by correlating tuning curves generated from odd and even trials within the same stage. The stability threshold was set at the 95th percentile of 1,000 randomly paired cell correlations.
d. Place fields along the linearised trajectory, sorted by place cell identity at each imaging stage. The left side of the white dotted line represents left-to-right trials; the right side represents right-to-left trials.
Data represent as mean ± s.e.m, Control; 10 imaging sessions from 5 mice per stage. 5xFAD: 7 imaging sessions from 4 mice per stage.

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Supplementary Figure 4. Correlation between hippocampal splitter cell quality and behavioural performance (related to Figure 4)
Pearson correlations between splitter-cell quality metrics and behavioural performance.
a. Splitting extent (top: per FOV; bottom: per mouse).
b. Trial-to-trial spatial correlation.
c. Discriminability between high- and low-rate trajectories.
d. Proportion of trials in which splitting activity was expressed.
e. Trial number at which splitting activity first emerged.
f. Decoding accuracy of a discriminant analysis classifier. Correlation coefficients and p values are shown in each panel. Data were pooled from 2–4 month-old and 7–12 month-old control (N=16) and 5xFAD mice (N = 13).

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78 **Supplementary Figure 5. Encoding on retrospective and prospective trajectories (related**
79 **to Figure 5)**
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a. Top: Dissimilarity values for retrospective trajectories (e.g., LR vs. LL trials; RL vs. RR trials).
Dissimilarity was computed as the $-\log_{10}$(p value) from Wilcoxon ranksum test comparing cell
activity rates at each spatial bin across the two trial types. Bottom: Retrospective dissimilarity
plotted across stem positions. Genotype effect: $F_{(15, 10809)} = 3.07$, $p < 0.001$; Genotype × bin
interaction: $F_{(15, 10809)} = 1.52$, $p = 0.089$; two-way ANOVA. Lower dissimilarity values indicate
stronger retrospective trajectory encoding.
b. Same analysis as in panel **a**, but for prospective trajectories (e.g., LR vs. RR trials; RL vs. LL
trials). Genotype effect: $F_{(15, 10904)} = 21.2$, $p < 0.001$; Genotype × bin interaction: $F_{(15, 10904)}$
$= 3.47$, $p < 0.001$; two-way ANOVA.

Supplementary Figure 6. Behaviour and hippocampal CA1 encoding after basal forebrain cholinergic activation via DREADDs

a. Number of trials completed after vehicle and CNO injection.

b. Average Trial speed after vehicle and CNO injection.

c-d. CA1 place cell metrics following vehicle and CNO injection (left to right): in-field versus out-of-field selectivity and peak in-field activity.

e. Place field size of CA1 place cells after vehicle and CNO injection.

f. Place cells information during correct trials after no i.p injection and vehicle i.p injection in DREADD (hM3Dq) injectied mice.

g-i. Stem place cell metrics following vehicle and CNO injection (left to right): split extent, trial-to-trial correlation in the high-rate trajectory direction, and the difference between high- and low-rate trajectory activity.

j. Representative examples of CA1 place cell activity after vehicle and CNO administration in a AD mouse. Magenta dots indicate cell calcium activity; grey lines indicate correct trials, and red lines indicate error trials. Cell ID 140 exhibited a place field on the reward arm, and Cell ID 304 showed a stem arm place field with splitting activity.

Data were analysed using linear mixed-effects models with DREADD activation (CNO vs vehicle) as a fixed effect and Mouse ID as a random intercept. Ctrl: 964 detected neurons; place cells: 697 (140 stem-place cells) after vehicle, and 522 (122 stem place cells) after CNO; 5 mice. AD: 777 detected neurons; place cells: 522 (46 stem place cells) after vehicle, and 449 (41 stem place cells) after CNO; 4 mice.