Synchronous Ensembles of Hippocampal CA1 Pyramidal Neurons Associated with Theta but not Ripple Oscillations During Novel Exploration

En-Li Chen[1,2], Tsai-Wen Chen[1,2], Eric R. Schreiter[3], Bei-Jung Lin[1,2*]

[1]Institute of Neuroscience, National Yang Ming Chiao Tung University, Taipei, Taiwan
[2]Brain Research Center, National Yang Ming Chiao Tung University, Taipei, Taiwan
[3]Janelia Research Campus, Howard Hughes Medical Institute, Ashburn, VA 20147, USA.
*Corresponding author: beijunglin@nycu.edu.tw

Abstract

Synchronous neuronal ensembles play a pivotal role in the consolidation of long-term memory in the hippocampus. However, their organization during the acquisition of spatial memory remains less clear. In this study, we used neuronal population voltage imaging to investigate the synchronization patterns of CA1 pyramidal neuronal ensembles during the exploration of a new environment, a critical phase for spatial memory acquisition. We found synchronous ensembles comprising approximately 40% of CA1 pyramidal neurons, firing simultaneously in brief windows (~25ms) during immobility and locomotion in novel exploration. Notably, these synchronous ensembles were not associated with ripple oscillations but were instead phase-locked to local field potential theta waves. Specifically, the subthreshold membrane potentials of neurons exhibited coherent theta oscillations with a depolarizing peak at the moment of synchrony. Among newly formed place cells, pairs with more robust synchronization during locomotion displayed more distinct place-specific activities. These findings underscore the role of synchronous ensembles in coordinating place cells of different place fields.

Introduction

The establishment of long-term memory involves the initial acquisition and subsequent consolidation of memory traces, processes intricately connected to the modification of synaptic transmission[1,2]. Synchronous ensembles, characterized by the simultaneous firing of multiple neurons over tens of milliseconds (referred to as population synchrony), play a pivotal role in influencing synaptic plasticity[3-5]. The hippocampus, a crucial region for long-term memory formation[6,7], has been a focal point of synchronous ensemble studies. Within the hippocampus, synchronous ensembles are hypothesized to play roles in reactivating and consolidating labile memory traces[8]. However, their organization during memory acquisition remains less understood.

In the hippocampus, synchronous ensembles among CA1 pyramidal cells (CA1PCs) are predominantly observed during sharp-wave ripples, the brain waves associated with memory consolidation[8-13]. During these events, 10-18% of CA1PCs generate spikes, particularly during offline consummatory behavioral states or sleep[14,15]. These synchronous ensembles are biased toward experience-dependent reactivation, consolidating labile memory traces[16,17]. Furthermore, synchronous ensembles firing action potentials in short windows of 10-30 milliseconds enhance information processing[6] and discriminate distinct behavioral contingencies[18]. Although synchronous ensembles of CA1PCs during offline memory consolidation are well-characterized, their organization during online acquisition of spatial memory, such as when animals enter a new environment and new place cells form, remains unclear.

Previous studies have identified correlated activities among place cells during theta oscillation, where 1-3% of place cells' spikes co-activate in a forward order matching animals' trajectory in the theta cycles when animals' locations overlap in the place fields of these cells[19-21]. Because these place cells fire at distinct theta phases in the same cycle, this coordinated firing operates on the timescale of theta periods (~125ms)[22,23]. Population synchrony on shorter timescales (10-30 ms) is

speculated to be reduced by the presence of theta oscillations[22].

The subthreshold membrane voltage (subVm), influenced by synaptic inputs and intrinsic

conductances, directly triggers spiking activity[24,25]. Additionally, properties of subVm support the

organization of neuronal representation[26]. Although correlated membrane potentials have been

proposed to underlie the synchrony of neuronal ensembles[27,28], multicellular recordings of subVms

have not yet provided direct experimental evidence. Given the intricate interaction among subVm,

spiking, and place cell formation, it is crucial to investigate the temporal relationships of sub- and

suprathreshold neuronal activities among CA1PCs during novel exploration.

In this study, we used voltage imaging to investigate the temporal dynamics of sub- and

suprathreshold neuronal activities in CA1PCs during novel exploration. We found synchronous

ensembles involving approximately 40% of CA1PCs exhibited transient population synchrony and

correlated subthreshold membrane potentials when mice ran and stopped in a new maze.

Surprisingly, these synchronous ensembles occurred outside ripple episodes and were phase-locked

to extracellular and intracellular theta oscillations. Notably, cell pairs with stronger synchronous

activities established more distinct place fields that overlapped less. In essence, synchronous

ensembles facilitated temporal association among place cells representing diverse features of

spatial memory. This plastic network, driven by synchronous ensembles, may contribute to the

stabilization of place cells, establishing place fields that effectively tile the environment.

Results

To investigate the temporal relationships between neuronal activities during novel exploration, we

simultaneously recorded many neurons as animals explored a new environment for the first time.

To achieve this, we employed voltage imaging in conjunction with an air-lifted plastic track that

allowed mice to explore an environment while remaining head-fixed under a microscope (Figure

1A). To ensure a genuinely novel experience, all pre-training and habituation were performed in a

separate room distinct from the one where the imaging experiments were conducted. On the day of

imaging, the mice were introduced to the recording room for the first time and initially confined to

a specific corner of the track (Figure 1A top view). Subsequently, as the imaging session

commenced, the confinement was lifted, granting mice the freedom to explore the track while

imaging was conducted.

In novel exploration, mice moved at a speed of 8.3±0.9 cm/s during locomotion (speed>3 cm/s) and

spent 53±7% and 36±6% of their time in locomotion and immobility (speed<1 cm/s), respectively

(n=5 mice). Meanwhile, they completed 10±2.5 laps during the 180-s imaging periods (n=5 mice).

For voltage imaging, we expressed the voltage indicator Voltron2[29] in CA1 pyramidal cells (CA1PCs),

allowing us to measure both suprathreshold and subthreshold membrane potentials from multiple

CA1PCs (14±4 cells imaged per field of view (mean±s.d.) and, in total, 71 cells imaged from 5 fields

of view in 5 mice; Figure 1B). Simultaneously, we monitored the animals' positions and speed within

the track while recording the local field potential (LFP) from the contralateral side of the

hippocampus.

The mean firing rates of cells ranged from 2.3 to 4.3 Hz (25th-75th percentiles) with a median of 3.2

Hz (n=71 cells, Supplementary Figure 1A). Additionally, most CA1PCs exhibited elevated mean firing

rates during immobility relative to locomotion, and their firing rates showed a negative correlation

with the animals' speed (Supplementary Figure 1B and 1C), consistent with previous research[30].

Synchronous ensembles among many CA1 pyramidal neurons

Figure 1C shows the fluorescence traces of all neurons in a session as an example. Instances of

synchronous activity, where many neurons emitted spikes within a narrow time window, were

frequently observed (Figure 1C). To investigate synchronous activity within the neuronal population,

we selected a reference cell and counted the number of spikes across all other cells that were

aligned with the spikes of the reference cell. This approach yielded a grand average cross-

correlogram (CCG) capable of elucidating the timing of spikes from the reference cell relative to the

combined spikes of the rest of the population. Notably, the grand average CCG exhibited heightened

spike counts around zero lags, unlike flattened counts calculated from randomly selected cells of

different sessions (Figure 1D and 1E). The full width at half maximum (FWHM) of the peak in the

grand average CCG measured 23±14 ms (mean±s.d., n=71 cells from 5 sessions; Figure 1F),

indicating the timescale of the population synchrony. Furthermore, pairwise CCGs of many cell pairs

showed significant peaks compared to jittered data (percentage of significant pairs: 63%, median,

n=497 pairs, Figure 1-figure supplement 1A and 1B). Among pairs with significant peaks, the

absolute values of the peak lags ranged from 0.8 to 4.3 ms (25th-75th percentiles) with a median of

2.1 ms, and the FWHM of the peaks ranged from 20 to 46 ms (25th-75th percentiles) with a median

of 30 ms (n=315 pairs, Figure 1-figure supplement 1C and 1D). These results highlight the

synchronous spiking between neurons on the millisecond timescale.

To systematically reveal individual synchronous events, we counted the number of spikes from all

simultaneously recorded neurons within sliding windows of 25 ms. To compare the summed spikes

with controls, we randomly jittered the spike trains to perturb spike timings while maintaining spike

rates. We identified moments of population synchrony when spike sums exceeded the average of

the jittered spike sums plus four standard deviations (Figure 1C). Using the same criterion for event

detection, we observed over a 7.8-fold increase in synchronous events in original data compared to

jittered controls (mean event rate: 0.94±0.17 Hz for the original data, 0.12±0.01 Hz for the jittered

data, n=10 segments, p=0.002, Wilcoxon signed-rank test; Figure 1G). Furthermore, we calculated

the percentage of neurons participating in these synchronous events to examine the synchronous

ensemble sizes. The ensemble sizes ranged from 32% to 53% (25th-75th percentiles) with a median

of 42%, which were larger than the synchronous ensemble sizes associated with ripples[14] (Figure

1H). These results demonstrate synchronous ensembles of CA1PCs engaging many neurons.

Figure 1 Population synchrony during novel exploration.

(A) Schematic of the behavioral setup for voltage imaging. Mice explored a novel one-way track while fluorescence images were captured at the CA1 pyramidal layer. Left: side view; right: top view of the setup.

(B) Example time-averaged image of CA1 pyramidal cells expressing Voltron2-ST labeled with JF_{552}-HaloTag ligand. Scale bar, 100 μm.

(C) Fluorescence traces of cells in an example session with identified synchronous ensembles (labeled with gray vertical bars). Right: an example of magnified fluorescence traces within the bracket, pinpointing the occurrence of the identified synchronous ensemble. Spikes are indicated by dots on the top.

(D) The grand average cross-correlogram (CCG) averaged across all 71 cells. The gray line represents the mean grand average CCG between reference cells and randomly selected cells from different sessions.

(E) Distribution of the grand average CCG peak lags of all cells (n=71).

(F) Distribution of the full width at half maximum (FWHM) of the grand average CCGs (n=71 cells).

(G) Pairwise comparison of the event rates of population synchrony between original and jittered data. Bar heights indicate group means. **p<0.01

(**H**) Histogram of the ensemble sizes as percentages of cells participating in the synchronous ensembles.

Synchronous ensembles during locomotion and immobility of novel exploration

The hippocampus exhibits state-dependent activities when rodents are engaged in locomotion and immobility[30,31]. In novel exploration, CA1PCs showed synchronous spiking during locomotion and immobility (Figure 2A). To investigate whether these synchronous activities differed between these behavioral states, we compared grant average CCGs and synchronous ensembles under these conditions. During both immobility and locomotion, grant average CCGs showed prominent peaks

compared to jittered data (Figure 2B). Moreover, synchronous ensembles were rich in both

immobility and locomotion periods (Figure 2A, 2B and 2D). Nevertheless, the FWHMs of grant

average CCGs were broader in the immobility group compared to the locomotion group (Figure 2B

and 2C). Synchronous events occurred more frequently during immobility compared to locomotion

(Figure 2D). Additionally, ensemble sizes were significantly larger in the immobility group compared

to the locomotion group (Figure 2E). Taken together, with differences in sizes and kinetics, there

were synchronous ensembles during both immobility and locomotion periods of novel exploration.

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Figure 2 Comparison of synchronous ensembles between immobility and locomotion.

(A) Example traces of cells in a session with identified synchronous ensembles (gray rectangles)

during periods of immobility (red rectangular frame) and locomotion (blue rectangular frame).

Spikes are indicated by dots at the top of individual traces. Top: the simultaneously recorded LFP

trace. Bottom: simultaneously recorded speed of the animal. Inset: the band-pass filtered LFP trace

at the theta frequency range (black line) overlaid on the raw LFP trace (gray line).

(B) The grand average cross-correlograms (CCGs) during immobility (red) and locomotion (blue)

averaged across all cells (n=71). The gray lines represent the mean grand average CCGs calculated

from jittered data.

(C) Pairwise comparison of peak widths (FWHM) in the grand average CCGs between immobility

(red) and locomotion (blue). Vertical bars represent means and standard deviations for both groups

(FWHM: 27$\pm$12 ms for immobility, 18$\pm$9 ms for locomotion, mean$\pm$s.d., $p<0.001$, paired Student's *t*-

test, n=71 cells). ***$p<0.001$

(D) Pairwise comparison of event rates of population synchrony during immobility and locomotion.

Bar heights indicate group means (immobility: 1.4$\pm$0.7 Hz, locomotion: 0.6$\pm$0.5 Hz, mean$\pm$s.d., n=10

segments, p=0.002, Wilcoxon signed-rank test). **$p<0.01$

(E) Boxplot of ensemble sizes for synchronous ensembles occurring during immobility (red) and

locomotion (blue) (median ensemble size: 47% for immobility, n=446 events, 32% for locomotion,

n=313 events, $p<0.001$, Wilcoxon rank-sum test). ***$p<0.001$

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Synchronous ensembles occur outside ripple episodes

CA1PCs exhibit synchronous activities when the hippocampal network emits ripple waves[8,14]. To test

whether there is a co-occurrence of synchronous ensembles and ripples, we first detected ripples

from LFP recordings. Ripples were detected in three of the five recording sessions, and all occurred

during immobility (mean ripple rate: 0.04$\pm$0.02 Hz during immobility, 0 Hz during locomotion). In

the two recording sessions where we did not detect any ripples, we verified the recording quality of

the LFP signal by running multiple tests after the recording sessions. Ripples were identified in the

same animals during quality tests, indicating that the absence of ripples during novel immobility in

the recording session was not due to the deterioration of LFP recordings (Figure 3-figure

supplement 1).

To visualize both ripples and synchronous ensembles, we plotted voltage traces of neurons

alongside simultaneously recorded ripple oscillations. It was typical for synchronous ensembles to

occur far away from ripple episodes (Figure 3A). To delve into the relationship between neuronal

activities and ripples, we aligned traces of ripple power with population synchrony and ripples. In a

recording session where 64 synchronous ensembles and 12 ripples were detected during

immobility, the LFP power at ripple frequency (120-240 Hz) exhibited no discernable peaks within

any of the synchronous events (Figure 3B, left upper and middle panels). Nevertheless, the same

LFP power displayed substantial peaks during every ripple event (Figure 3B, right upper and middle

panels). A striking contrast between synchronous events and ripples was also evident when spiking

probabilities during these events were examined. During population synchrony, the mean spiking

probability averaged across all cells was high, while the spiking probability was almost zero during

ripple events (Figure 3B lower panel). In the three sessions where ripples were identified, we

calculated the percentages of ripple events coinciding with synchronous ensembles and vice versa.

Remarkably, there was no co-occurrence of ripples and synchronous ensembles (Figure 3C).

Furthermore, we quantified ripple-modulated spiking by calculating the ratio of the difference in

mean firing rates during in-ripple and out-ripple periods to the sum of the mean firing rates during

both periods. Indexes of most cells were negative, and a striking 85% of cells displayed ripple

modulation indexes of -1, indicating complete suppression of their spiking activities during ripples

(Figure 3D). Taken together, during novel exploration, synchronous ensembles involved many

neurons with millisecond-synchronized spikes and occurred outside the time frames of ripples.

Figure 3 Synchronous ensembles occur outside ripple episodes.

(A) Example traces with ripples and synchronous ensembles. Top: simultaneously recorded LFP

displaying an episode of ripple oscillation. A red vertical line indicates the timing of the peak ripple

power in the LFP trace. A yellow rectangle marks the duration of the ripple episode. Inset at the top:

a magnified view of the ripple oscillation. Middle: fluorescence traces of 9 CA1PCs. Gray rectangles

mark the timings of the synchronous ensembles on all traces. Bottom: animals' speed recorded

simultaneously with voltage imaging.

(B) LFP ripple power and spiking probability aligned to synchrony ensembles and ripple events.

Upper panel: color-coded LFP power at ripple frequency (120-240Hz) aligned to the timings of 64

synchronous ensembles (left) and 12 ripple events (right). Middle panel: Average LFP power of

ripple frequency aligned to the timings of the synchronous ensembles (left) and the ripple events

(right). Lower panel: Average spike counts of all cells aligned to the timings of the synchronous

ensembles (left) and ripple events (right).

(C) Percentages of co-occurrences between population synchrony and ripples. The co-occurrences

in the "sync to ripples" group represent the percentages of ripple episodes co-occurring with

synchronous ensembles in the same episodes. The co-occurrences in the "ripples to sync" group

indicate the percentages of synchronous ensembles co-occurring with ripple episodes. Each gray

circle represents an individual session.

(D) Distribution of the ripple modulation indexes (n=41 cells). Gray bar: modulation indexes with

significant p values (n=35 cells); white bars: modulation indexes without significance (n=6 cells).

Synchronous ensembles are associated with theta oscillations

To investigate the network dynamics associated with synchronous ensembles, we aligned LFP traces

triggered by the timings of synchronous ensembles. The average LFP traces of a representative

session showed prominent theta oscillations around the timings of both immobility and locomotion

population synchrony (Figure 4A). Spectral analysis of the mean synchrony-triggered LFP traces

during immobility and locomotion revealed theta frequency components of 4-12 Hz, with a slightly

higher peak frequency in locomotion LFP compared to immobility LFP (Figure 4B). Furthermore, we

computed LFP theta modulation and identified preferred phases of synchronous ensembles. In line

with the previous result, synchronous ensembles displayed a high level of modulation by LFP theta

oscillation (modulation strength: 0.61±0.05 during immobility and 0.74±0.04 during locomotion, n=5 sessions). Although the preferred phases varied from session to session due to differences in recording sites along the proximal-distal axis of the hippocampus, the timings of individual ensembles were consistently locked to the preferred phase of each session (Figure 4C). These results establish a strong link between synchronous ensembles and LFP theta oscillation.

Subthreshold membrane voltage (subVm) is one of the critical parameters supporting the generation of action potentials. To investigate the subthreshold signals underlying population synchrony, we aligned subVm traces with the timings of synchronous ensembles and observed that both immobility and locomotion synchrony rode on a subVm waveform oscillating at theta frequency (Figure 4D). To elucidate further the relationship between subVm theta oscillation and spikes involved in population synchrony, we calculated the subVm theta modulation and preferred phases of spikes. Spikes participating in both immobility and locomotion synchrony were strongly coupled to the subVm theta oscillation of neurons, with the mean preferred phase being in the latter half of the rising phase closer to subVm theta peaks (Figure 4E). Conversely, spikes not participating in synchronous ensembles had weaker modulation strength compared to spikes involved in synchrony (Figure 4-figure supplement 1A and 1B). Furthermore, the membrane voltages triggered by these two categories of spikes showed a significant difference in the hyperpolarization phase of the theta oscillation. Synchronous spikes were situated within cycles of theta oscillations featuring more hyperpolarized troughs than spikes outside synchronous ensembles (Figure 4-figure supplement 1C and 1D). Thus, compared to spikes not participating in synchronous ensembles, synchronous spikes showed a more robust modulation by theta oscillation of membrane voltage during immobility and locomotion in the context of novel exploration.

Notably, phase locking of spikes to their subVm theta oscillation and the synchronous spikes among many neurons suggest that subVms of different neurons would be correlated at the theta frequency.

To explore the possibility, we divided the 180-second subVm traces into 1-second segments with

50% overlap and calculated the cross-correlation and magnitude-squared coherence function for

each pair of cell segments. Segments were sorted into the immobility or locomotion group based on

the animals' speed during each segment (kept below 1 cm/s for immobility and above 3 cm/s for

locomotion for more than 90% of the time). In total, we analyzed 511 cell pairs from 5 sessions.

Indeed, the cross-correlation of subVm between cell pairs revealed a high correlation at zero lag and

periodic modulation of the correlation at theta frequency (Figure 4F). Furthermore, the subVm

theta coherences were increased in both groups (immobility: 0.50, locomotion: 0.40, median,

n=511 pairs) and were negatively correlated with soma distances (immobility: -0.46, locomotion: -

0.48, n=511 pairs, Figure 4G). Taken together, during novel exploration, CA1PCs generate

synchronous activity associated with theta oscillation and exhibit correlated and theta-coherent

subVms.

Figure 4 Theta oscillations in the LFP and subthreshold membrane potential were linked to

population synchrony.

(A) LFP waveforms triggered by timings of synchronous ensembles. Red: triggered by timings of

immobility synchrony; blue: triggered by timings of locomotion synchrony.

(B) Power spectral densities of the triggered LFP waveforms. Red: immobility; blue: locomotion.

(C) Distributions of the differences between LFP theta phases of individual synchronous ensembles

and their preferred phases of the session. Note that most ensembles show minor differences within

±30 degrees. Red: immobility; blue: locomotion.

(D) Fluorescence traces triggered by timings of synchronous ensembles. Traces with spikes are

excluded. Left panel: traces triggered by timings of immobility synchrony. Right panel: traces

triggered by timings of locomotion synchrony. Thin gray lines: mean triggered fluorescence traces

averaged across triggers for representative cells. Thick red and blue lines: mean triggered

fluorescence traces averaged across cells.

(E) Polar plot illustrating theta modulation of spikes participating in the immobility synchrony (red

dots) and of spikes participating in the locomotion synchrony (blue dots). The angle of the dots

indicates the preferred phase, and the distance from the center of the circle represents the

modulation strength. Each dot represents the averages from a neuron.

(F) Pairwise cross-correlation of the subthreshold membrane voltages (subVm). Top: color-coded

cross-correlation of subVm between cell pairs during immobility and locomotion periods. Bottom:

averaged cross-correlation over all cell pairs for immobility (red) and locomotion (blue) subVm. Both

show a central peak at zero lag flanked by theta oscillation.

(G) Scatter plots of theta coherence against soma distances of immobility subVm (red) and

locomotion subVm (blue) between cell pairs.

Synchrony between place cells with distinct place tunings

When animals explore a novel maze, spatially tuned spiking activity can form within a few minutes[32,33]. However, how synchronous ensembles organize their spatially tuned activities during novel exploration remains to be determined. Synchronous ensembles, generating spikes coincidently in short intervals, might display correlated spiking at longer time windows that match behavioral timescales of exploration, potentially leading to shared place responses. On the other hand, CA1PCs are interconnected through inhibitory interneurons[34-36]. Modulation of activities in inhibitory interneurons by some place cells may influence the activities of other place cells, resulting in divergent spatial tunings[32,37,38].

To compare the spatial tunings of synchronous cell pairs, we first calculated the spatial tuning curves of individual neurons by computing the mean firing rates in every spatial bin. On average, 40% of cells displayed selective spatial tunings of their spiking activities and were qualified as place cells[39] (Figure 5A). Among place cell pairs, some showed similar spatial tunings, while others did not (Figure 5B). For instance, in an illustrative session depicted in Figure 5B, the seventh and eighth place cells had highly similar spatial tunings with a correlation coefficient as high as 0.7. In contrast, the fifth and tenth place cells exhibited vastly different spatial tunings with a correlation coefficient as low as -0.5.

We further explored the relationship between the similarity of spatial tunings and synchronization strength among pairs of place cells. We measured synchronization strength as the peak of the CCG normalized by the averaged jittered spike count within the same bin as the CCG peak. We segregated spikes based on these behavioral states to accommodate differences between locomotion and immobility and calculated the corresponding CCG and synchronization strength.

During locomotion, we observed a robust negative correlation between synchronization strength

and the similarity of spatial tunings among cell pairs. In other words, cell pairs with more precisely

co-active spikes during locomotion exhibited more distinct place-tuning profiles (Figure 5C and 5D).

This anti-correlation between synchronization strength and place tunings indicates coordination

between place cells' temporal and spatial coding. On the other hand, synchronization strength

during immobility showed little correlation with the similarity of spatial tunings among place cell

pairs, suggesting that correlated activities at the millisecond timescale do not extend to slow,

behavioral-relevant timescales (Figure 5-figure supplement 1). Thus, place cells with distinct place

fields are linked by synchronous activity during novel locomotion.

Figure 5 Negative correlation between spatial tuning similarity and synchronization strength

during novel exploration.

(A) Spatially binned rate maps of place cells. High firing rates are indicated in yellow, and low firing

rates are noted in blue. Bottom: the mean firing rates along spatial bins averaged across laps.

(B) Spatial tunings of 10 place cells from an example session. Cells are sorted based on the locations

of their peak firing rates.

(C) Two example pairs of neurons with cross correlograms during locomotion and spatially binned

firing rate maps.

(D) A scatter plot of correlation coefficients of spatial tunings against synchronization strengths for

place cell pairs. Each dot represents a pair of place cells. Synchronization strengths are negatively

correlated with similarities of spatial tunings (Spearman correlation coefficient=-0.4, p<0.001, n=73

pairs).

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Discussion

Our findings reveal the presence of synchronous ensembles comprising a substantial number of

CA1PCs during both locomotion and immobility in a novel environment. These ensembles

concurrently engage neurons representing diverse aspects of the environment and are rarely

observed during ripple events. Instead, they closely align with LFP and intracellular theta

oscillations. Our data demonstrate theta-associated synchronous ensembles during spatial memory

acquisition, coordinating numerous CA1 place cells that transmit information about different

segments of the environment.

The correlation between observed population synchrony and theta oscillations is intriguing. Theta

oscillations are typically associated with the sequential activation of neurons over time, resulting in

reduced neuronal synchrony[22]. However, theta oscillations come in different types, each

characterized by distinct pharmacological properties and behavioral correlates[40-42]. For example,

type 1 theta, referred to as translational theta, occurs during voluntary movements and is sensitive

to N-methyl-D-aspartate receptor (NMDAR) blockers. On the other hand, type 2 theta, known as

attentional theta, is blocked by muscarinic receptor antagonists, emerging during states of arousal

or attention, such as entering a new environment[43-45].

During theta oscillations, cholinergic neurons in the medial septum excite downstream CA1PCs,

leading to increased firing rates and membrane potentials[46-48]. Given the widespread influence of

cholinergic inputs on numerous CA1PCs, theta-coupled cholinergic inputs may drive depolarization

and firing across many CA1PCs. Additionally, they could enhance the likelihood of synchronicity.

Furthermore, GABAergic neurons in the medial septum, activated during theta oscillations, could

activate interneurons in the CA1 region[42,49,50]. These CA1 interneurons, in turn, have the capability

to entrain the activities of pyramidal neurons at both sub- and suprathreshold levels, thereby

promoting synchronization among pyramidal neurons[51,52]. In summary, cholinergic and GABAergic

neuron activation during theta waves may contribute to the excitation of CA1PCs and the

consequent increase in their synchronicity.

Theta oscillations have long been implicated in memory acquisition. Specifically, theta power

correlates positively with learning performance[53-55]. Perturbing generation of theta waves leads to

impairment in memory acquisition[56-60], while electrical stimulation restoring theta rhythms rescues

learning performance[61,62]. Furthermore, increased spike-theta coherence correlates with better

memory, and stimulations that boost theta phase-locking of spikes enhance memory[63-65]. Despite

abundant evidence supporting a link between theta oscillations and memory, the precise network

mechanism by which theta oscillations support memory acquisition remains unclear. Theta

oscillations are hypothesized to link spatially distributed neurons into functional ensembles to

support memory acquisition[13]. The identified synchronous ensembles support the hypothesis,

providing a network substrate that links theta rhythms to the initial stage of memory acquisition.

Previous reports of CA1 synchronous ensembles occur predominantly during ripple

oscillations[8,14,17]. However, we found only a few ripple events in the initial minutes of entering a

new environment, even when animals are not moving. During these events, CA1PCs show

decreased spiking activities, and they display synchronous firing outside ripple episodes (Fig.3).

Different from the ripple-associated synchronous ensembles described previously, the synchronous

ensembles we observed involve larger populations of neurons spiking synchronously in narrower

time intervals and are associated with theta oscillations. The absence of recorded ripple-associated

synchrony in our study may be attributed to elevated arousal with prominent theta oscillations even

during immobility, potentially suppressing ripples[66]. Additionally, the deep-layer CA1PCs imaged in

our study are known to show decreased activities during ripples, contributing to their lack of ripple-

associated synchrony[67]. The CA1 network may generate multiple types of population synchrony,

with ripple-associated synchrony facilitating memory consolidation, while the newly identified theta

population synchrony supports the initial stage of memory acquisition.

Although CA1PCs are known to exhibit weaker recurrent connectivity than that of CA3 pyramidal

neurons, CA1PCs did communicate through monosynaptic connections[68]. These connections can be

dynamically adjusted by the spike-timing-dependent plasticity mechanism[69-71], and therefore could

be affected by the transient population synchrony. In addition, CA1PCs are known to communicate

through inhibitory interneurons. Multiple pyramidal neurons converge onto a single interneuron,

and individual connections are often weak and unreliable[35,72,73]. This connectivity scheme requires

multiple pyramidal neurons spiking synchronously within the time window of synaptic integration to

effectively activate postsynaptic neurons[35]. Furthermore, population synchrony among CA1PCs may

propagate to downstream areas, synchronizing neurons outside the hippocampus. With their large

ensemble sizes, these synchronous ensembles may synergistically convey relevant information or

reorganize the connectivity of the neuronal networks.

Prior studies in sensory systems have commonly revealed that neurons with synchronous activities

typically share similar sensory responses[74,75]. However, the synchronous ensembles we have

identified often encompass neurons with distinct preferences for place fields. Pairwise synchrony

demonstrates a negative correlation with place field overlap (Fig. 5). These findings indicate that

synchrony does not necessarily imply similar tunings between neurons. Furthermore, these

synchronous ensembles involve place cells encoding different spatial information. Recognizing the

crucial role of neuronal synchrony in binding distributed neurons in the brain[3,5], population

synchrony may serve as a mechanism to integrate diverse features of the same environment into a

cohesive mental map.

Methods

Animal

Time-pregnant wild-type C57BL/6J female mice underwent in utero virus injection of CamkII-cre virus (105558-AAV1, Addgene) into the left side of the lateral ventricles. The resulting offspring were raised in a breeding room under controlled temperature and humidity conditions and a 12-hour light/dark cycle (lights on from 7:00 to 19:00). The mice used in the experiments were between 12 and 18 weeks old at the time of the recordings. All surgical procedures, behavioral training, and recording protocols were approved by the NYCU Institutional Animal Care and Use Committee.

In-utero virus injection

Time-pregnant mice on an embryonic day 14.5 were anesthetized with isoflurane (induction:4-5%; maintenance:1-3%) before surgery. The animals were placed on a heating pad in a supine position to maintain body temperature. Aseptic procedures were implemented to maintain sterile conditions during the surgery. Uterine horns were exposed one at a time using spoons and fingers. Warm and sterile phosphate-buffered saline (PBS) was used to rinse the uterus to prevent it from drying out. Micropipettes loaded with CamkII-cre virus solution were used to inject 0.2 uL of virus suspension into the left side of the lateral ventricle of each embryo. The uterine horn was then gently returned to the abdominal cavity. The exact process was repeated to the other uterine horn. After the surgery, ketorolac (6mg/kg) and cefazolin (1g/kg) were administered for two days to minimize pain and inflammation.

Cranial window and cannula implantation

The offspring of the mice that had undergone in-utero injection were subjected to chronic hippocampal window surgery when they were between 2 and 4 months old. The animals were

initially anesthetized using isoflurane (induction:4-5%; maintenance:1-3%). Topical anesthesia was

applied by administering 0.5% lidocaine to the wound margins. After achieving deep anesthesia, an

incision was made in the skin, and a circular section of the skull was removed, centered 2 mm

caudal and 1.8 mm left to the bregma. Using 30-gauge needles and forceps, the dura was removed

from the exposed area, and cortical tissue within the craniotomy was aspirated. A fine injection

pipette (tip diameter 10-60 um) was used to inject AAV2/1-CAG-flex-Voltron2-ST (2.7×10^{12} GC/ml, a

courtesy gift from Dr. Eric Schreiter and the GENIE team at HHMI Janelia Research Campus) into the

exposed regions at a depth of 200 μ m (up to six injection sites and 100-200 nL of viral

suspension). An optical chamber was constructed by placing a cannula with a cover slip attached at

the bottom over the craniotomy and sealing it with dental acrylic or C&B Metabond (Sun Medical).

A custom-built titanium frame was then cemented to the animal's head using dental acrylic or C&B

Metabond. Mice received ketorolac (6mg/kg) and cefazolin (1g/kg) for two days following surgery to

minimize pain and inflammation.

Behavioral training

After at least two weeks of recovery from the window surgery, the mice underwent behavioral

training to ensure they were calm and attentive in the test environment. The initial training phase

involved acclimating the mice to head fixation and treadmill running, which took place over 3-5 days

in a separate room distinct from the recording setup. Following this training, the imaging

experiments were conducted. On the day designated for imaging, the mice were introduced to the

recording room for the first time just before the imaging session. During the imaging experiments,

the mice were securely held in a head-fixed position beneath the microscope and on an air-lifted

plastic track that rested on an air table (AirLift, Neurotar; Fig. 1Aa). The track was adorned with

various shapes of signs to help the mice navigate the track. Subdued blue lighting was provided to

illuminate the track. Before the initiation of imaging, the mice were confined in a corner of the track

using gates. Subsequently, the gates were opened upon the start of imaging, allowing the mice to

explore the new environment during their initial laps (Fig. 1Ab). The Mobile HomeCage tracking

system (Neurotar) was used to track the mice's position and speed. During the acquisition of image

frames, TTL pulses were sent by the high-speed camera to the Mobile HomeCage tracking system to

facilitate data alignment.

Voltage imaging

Imaging was performed after treadmill training. JF_{552}-HaloTag ligand[76] (a courtesy gift from Dr. Luke

Lavis) was first dissolved in DMSO (Sigma) and then diluted in Pluronic^{TM} F-127 (P3000MP,

Invitrogen) and PBS to achieve a final concentration of 0.83 mM of JF_{552}-HaloTag ligand. The

solution was then injected intravenously through the retro-orbital sinus. Imaging sessions were

initiated 3-5 hours after the injection of the JF_{552}-HaloTag ligand. Fluorescence sensors were excited

using a 532-nm laser (Opus 532, Laser Quantum) with an excitation filter (FF02-520-28, Semrock),

and the emitted light was separated from the excitation light using a dichroic mirror (540lpxr

Chroma) and passed through an emission filter (FF01-596183 Semrock). A 16X, 0.8 NA objective

(Nikon) was used to collect the emission light which was then imaged on a CMOS camera (DaVinci-

1K, RedShirt) with a 50-mm camera lens (Nikkor 50mm f1.2, Nikon) as the tube lens. Images were

acquired using Turbo-SM64 (Sci-Measure) at 2kHz with a resolution of 190 x 160 pixels,

corresponding to a field of view of 1.4 x 1.2 mm. The number of frames in an image session was set

at 360,000, corresponding to a duration of 180 seconds per session. Imaging depths ranged

between 70 and 170 μm from the window surface, where the CA1 pyramidal neurons were

located. Time-lapse images were collected, and image stacks of the same field of view along the z-

axis were acquired. The image stacks consisted of 100 plans separated by two μm, covering the

depth from the window surface to 200 um from the surface.

LFP recording

The LFP signal from the contralateral CA1 region of the hippocampus was simultaneously recorded

during voltage imaging. To implant the LFP electrode, a small craniotomy was performed at the

dorsal CA1 coordinate (2mm caudal and 1.8 mm right to the bregma). A tungsten electrode

(#100211, 38 μ m in diameter, polyimide insulated, California Fine Wire) was inserted into the

dorsal CA1 region, and the wire and pins were cemented in place with dental acrylic. The electrical

signal was amplified and filtered (1-1kHz) using a DAM80 amplifier (Word Precision Instruments).

The amplified signal was acquired by an I/O device (National Instruments USB-6341) and recorded

using WaveSurfer (Adam Taylor, HHMI Janelia Research Campus) at 10 kHz during implantation and

TurboSM64 (RedShirt) at 8 kHz during imaging.

Voltage imaging preprocessing and spike detection

Fluorescence intensities were corrected for brain movement using rigid registration. Regions of

interest (ROIs) were manually selected by grouping pixels that cover individual somata. The

fluorescence intensities of individual neurons were calculated by averaging the fluorescence

intensities of pixels from the same ROIs. Bleaching was corrected by calculating the baseline

fluorescence (F_0) at each time point as an average of the fluorescence intensities within $\pm$0.5 s

around the time point. The dF/F was calculated as the F_0 minus the fluorescence intensity of the

same time point divided by F_0. Positive fluorescence transients were detected to identify spikes

from the high-passed dF/F traces created by subtracting the dF/F traces from the median-filtered

version with a 5-ms window. To simulate the noise of recordings, high-passed dF/F traces were

inverted, and the amplitudes of the transients detected from the inverted traces were used to

construct a noise distribution of the spike amplitudes. A threshold was set by comparing the

amplitudes of the detected transients with the noise distribution of the spike amplitudes to

minimize the sum of type I and type II errors. Spikes were first detected when transients were larger

than the threshold. Then, spike amplitudes smaller than half of the top 5% spike amplitudes were

excluded. The signal-to-noise ratio (SNR) was calculated for each neuron as a ratio of the averaged

spike amplitudes over the standard deviation of the high-passed dF/F traces, excluding points 2 ms

before and 4 ms after each detected spike to estimate the quality of the recordings. The crosstalk

between ROIs was routinely checked by examining the refractory periods of neurons in auto-

correlograms. Only neurons meeting the following criteria were included: (1) SNR larger than 5, (2)

full width at half maximum of the spike waveform longer than 0.8 ms, (3) mean spike rates higher

than 0.1 Hz, (4) distances between soma pairs at least 70 μm. Sessions with at least nine neurons

meeting the selection criteria were used for further analysis. Based on these criteria, we included

five sessions containing 9, 11, 13, 19, and 19 neurons.

Behavioral states of locomotion and immobility

The periods of locomotion were defined as instances when the animals' speed exceeded 3 cm/s,

while periods of immobility were determined as instances when the animals' speed fell below 1

cm/s.

Mean firing rates during locomotion and immobility

To calculate the mean firing rates during locomotion, we counted the number of spikes during

locomotion periods and divided this number by the total duration of locomotion. Similarly, for mean

firing rates during immobility, we calculated the number of spikes that occurred during immobility

periods and divided this number by the total duration of immobility.

Correlation between instantaneous firing rate and animal speed

To estimate the instantaneous firing rates of individual neurons over time, we convolved their spike

trains with a Gaussian window of 250 ms. Subsequently, we calculated the Pearson correlation

coefficient between these instantaneous firing rates and the animal's speed for each neuron.

Grand average cross-correlogram (CCG)

Grand average CCGs were generated by constructing histograms of relative spike timings between

spikes of a reference cell and spikes of all other neurons in the same session as the target cell

population. Histograms were binned using 1-ms time bins. Spike counts were normalized by the

number of reference spikes times the number of all other neurons. An equal number of cells from

different sessions were randomly selected to form the shuffled target cell population to assess the

statistical significance of the CCG peaks. Subsequently, grand average CCGs were computed using

shuffled spike trains for comparison. Grand average CCGs were calculated using all spikes in the 180-

s recording periods, subsets of spikes during immobility periods (immobility grand average CCGs),

and subsets of spikes during locomotion periods (locomotion grand average CCGs).

Detection of the population synchrony

To assess population synchrony, we counted the number of spikes from all neurons within a sliding

window of 25 ms. To test the significance of the synchrony, we generated surrogate data by jittering

spike timings within a $\pm$75-ms window. By comparing the spike counts between the original and

jittered populations, we determined population synchrony when the spike count was higher than

the mean plus four standard deviations of the jittered spike counts. To estimate synchronous event

rates, we divided the entire 180-s recordings into halves and calculated the event rate in each of the

90-s segments. Additionally, we quantified the ensemble sizes by counting the percentage of

participating neurons in each synchronous event.

Pairwise cross-correlogram (CCG) of neuronal pairs

To compute the CCG, we first determined the time difference between spikes in the target and

reference neurons and then binned these values into 1-ms time bins. CCGs were only computed for

pairs with more than 100 counts within the CCG window of $\pm$500 ms to ensure reliable correlation.

We defined the peak of the CCG as the local maxima within $\pm$30 ms of lag. To assess the significance

of cross-correlations, we generated surrogate data by jittering the spike times of the target neuron

within a ±75 ms window. We computed the CCG peak for a thousand iterations and calculated p-

values by comparing the percentile of the peak in the null distribution of the jittered peak values.

The synchronization strength was estimated as the ratio between the original CCG peak and the

mean spike count of the same bin as the original CCG peak averaged across the jittered histograms.

For display purposes, the CCGs were smoothed using a 5-point moving average. CCGs were

calculated using all spikes in the 180-s recording periods, subsets of spikes during immobility

periods (immobility pairwise CCGs), and subsets of spikes during locomotion periods (locomotion

pairwise CCGs).

LFP ripple analysis

To detect ripples, we performed band-pass filtering of the LFP signal between 120 and 240 Hz,

followed by computing the envelope using the absolute values of the Hilbert transform, which was

then low-pass filtered at 20 Hz. The start and end times of ripples were identified by setting the

upper and lower thresholds at the mean plus 7 and 3.5 times the standard deviation of the

envelope, respectively. For a ripple event to be included, three conditions had to be met: (1) values

within the event were higher than the lower threshold, (2) at least one value exceeded the upper

threshold, and (3) the minimum duration of the event was 30 ms. The start and end times of a

ripple event were defined as the positive and negative crossings of the lower threshold,

respectively. We defined in-ripple periods as the time between the start and end of detected LFP

ripple events, while out-ripple periods were periods outside the in-ripple periods. The envelope

peak values of the detected ripple events were used as the time of ripple occurrence. These timings

were utilized as trigger timings to calculate averages triggered by ripples. Two co-occurrences were

quantified: the percentages of LFP ripple oscillations that occurred with synchronous ensembles

and the percentages of synchronous ensembles that occurred with LFP ripple oscillations.

Ripple modulation index

We utilized the ripple modulation index to assess changes in firing rates during ripples compared to

periods outside ripples. This index was computed as the ratio of the difference in mean firing rates

during in- and out-ripple periods to the sum of the mean firing rates during both periods.

Specifically, we first calculated the mean firing rate during in-ripple periods and the mean firing rate

during out-ripple periods. Next, we subtracted the mean firing rate during out-ripple periods from

the mean firing rate during in-ripple periods and divided the result by the sum of the mean firing

rates during both periods. The resulting ripple modulation index indicates the magnitude and

direction of the firing rate modulation during ripple compared to periods outside ripples.

LFP traces triggered by population synchrony and analysis

The midpoints of the synchronous event intervals were utilized as triggers to align LFP traces. These

traces were aligned within a ± 1-second window around the trigger time points. To calculate the

spectral density functions of the aligned LFP traces, we performed a 16,384-point fast Fourier

transform of the 2-second LFP traces. The power spectra were then estimated as the squared

absolute values of the Fourier coefficients. Ripple power was calculated as the sum of the power

distributed between 120 and 240 Hz. To calculate theta modulation strength and preferred angles of

the synchronous ensembles, LFP was first filtered within the theta frequency range. Then, the phase

$\varphi(t)$ and the instantaneous amplitude $A(t)$ of the filtered LFP were computed using the Hilbert

transform. A vector $V_k = A(t_k)e^{i\varphi(t_k)}$ was assigned to each synchronous ensemble that occurred

at time t_k. The modulation strength and the preferred phase angle were determined as the

absolute value and the phase angle of the summed vector over all ensembles, respectively.

Modulation strengths were normalized by the total length of all vectors.

Subthreshold membrane voltage (subVm)

To compute the subVm of individual neurons, we began by determining the difference between the

raw dF/F signal and the high-pass version of dF/F. Next, we calculated the spike threshold for each

cell by identifying the voltage value corresponding to the peak of the first derivative of the spike

waveforms. Any fluorescence values in the slow dF/F time courses that surpassed the spike

thresholds were linearly interpolated.

SubVm traces triggered by population synchrony

The midpoint of the time intervals, during which synchronous ensembles were detected, served as

the reference points for aligning fluorescence traces within a $\pm$300 ms window around those

reference points. To eliminate any interference from membrane potential fluctuations associated

with spikes, we excluded aligned traces in cases where any spike occurred within the time intervals

of the aligned traces.

SubVm theta modulation and phase preference of spikes occurred inside and outside the

population synchrony events

The subVm was initially filtered within the theta frequency range. Subsequently, the phase $\varphi(t)$

and the instantaneous amplitude $A(t)$ of the filtered subVm were computed using the Hilbert

transform. Spikes were categorized into four groups: those occurring during or outside immobility

synchronous events and those occurring during or outside locomotion synchronous events. Within

each category, a vector $V_k = A(t_k)e^{i\varphi(t_k)}$ was assigned to each spike that occurred at time t_k.

The modulation strength and the preferred phase angle were determined as the absolute value and

the phase angle of the summed vector over all spikes, respectively. Modulation strengths were

normalized by the total length of all vectors to facilitate comparison between neurons.

Spike-triggered fluorescence traces

We extracted segments of fluorescence traces using a $\pm$300 ms time window centered on the spike

timings. To examine variations in fluorescence waveforms triggered by spikes within and outside

synchronous events, we categorized the fluorescence traces based on whether the spikes occurred

within or outside these events. Subsequently, we performed pairwise comparisons of the

fluorescence values from the same neuron, concentrating on spikes occurring during corresponding

behavioral states. Neurons with fewer than four triggering events in any of these categories were

omitted from the analysis.

Cross-correlation and coherence analysis to the subVm traces

To calculate cross-correlation and theta coherence of subVm between pairs of neurons during

locomotion and immobility, we initially divided the 180-second subVm traces into 1-second

segments with a 50% overlap. These segments were then categorized into the locomotion group if

the animal's speed remained above three cm/s for more than 90% of the time within that segment.

Similarly, segments were sorted into the immobility group if the animal's speed remained below

one cm/s for more than 90% of the time within that segment. Subsequently, we computed the

cross-correlation and magnitude-squared coherence function of frequency for pairs of co-occurred

segments belonging to a specific cell pair. Mean cross-correlations of individual cell pairs were

averaged across segments that occurred in the same behavioral states (i.e., locomotion or

immobility states). To estimate theta coherence, we averaged the values of the coherence function

within the theta frequency range (4-12 Hz). Theta coherences from the same cell pair during the

same behavioral state were pooled together, and the average was used as an estimate for the theta

coherence of that cell pair during a particular behavioral state. Finally, Spearman correlation

coefficients were calculated between the mean theta coherences of cell pairs and their soma

distances.

Spatial analysis

We partitioned a 90-cm-long track into 36 spatial bins to analyze spatial tuning properties. For each

cell, the spatial tuning curve was computed by determining the mean firing rate, defined as the

number of spikes within the bin when the animal's speed exceeded three cm/s, divided by the total

dwell times using the same speed threshold. The peak firing rates were identified as the maximal

rate on the spatial tuning curve. Spatial selectivity was calculated as one minus the circular variance

of the spatial tuning curve. Neurons were classified as place cells when their spatial selectivity

exceeded 0.25, and peak firing rates exceeded 1 Hz. To assess the similarity of spatial tuning

between pairs of neurons, Spearman correlation coefficients were calculated for the spatial tuning

curves. To explore the relationship between levels of synchronization and spatial tuning similarity

among pairs of place cells, Spearman correlation coefficients were computed between

synchronization strengths of cell pairs and correlation coefficients of their spatial tuning curves.

Data analysis and statistics

Data analysis was performed using MATLAB (Mathworks, R2019b), and results were presented as

standard error of the mean unless otherwise specified. Boxplots were used to depict the median,

first quartile, and third quartile, with the middle line, the upper edge, and the lower edge of the

box, respectively. The whiskers of the boxplots represented the 5th and 95th percentiles of the

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Supplementary Figure 1 Mean firing rate distribution and correlation with animals' speed.

(A) Cumulative distribution of mean firing rates averaged throughout the entire 180-second

recording period.

(B) Mean firing rates during locomotion plotted against mean firing rates during immobility for

individual cells. Each dot represents a cell. Mean firing rates are higher during immobility than

during locomotion (4.2±0.2 Hz for immobility, 2.8±0.2 Hz for locomotion, n=71 cells, p<0.001,

Wilcoxon signed-rank test).

(C) Histogram of correlation coefficients between instantaneous firing rates and the animals' speed

for each neuron.

Figure 1-figure supplement 1 Pairwise cross-correlogram (CCG) between CA1 pyramidal cells.

(A) CCGs of two representative cell pairs exhibiting significant peaks compared to the jittered spike

trains of the target cells (gray lines).

(B) Distribution of the CCG peaks (n=497 pairs). Black: CCG peaks with significance (p<0.05). White:

CCG peaks without significance (p>=0.05).

(C) Histogram of peak lags of significant CCG peaks (n=315 pairs).

(D) Histogram of FWHMs of significant CCG peaks (n=315 pairs).

Figure 3-figure supplement 1 LFP ripples recorded in the first and subsequent sessions in a novel

maze. (A) An example LFP trace with ripples recorded in a subsequent session following the first

session of novel exploration, where no ripple was detected.

(B) Ripple rates in the first and subsequent sessions recorded in the same animals (n=5 mice) during

novel exploration.

Figure 4-figure supplement 1 Theta modulation of spikes during immobility and locomotion

periods of novel exploration.

(A) Polar plot comparing theta modulation between spikes participating in synchronous ensembles

(sync spikes) and spikes not participating in synchronous ensembles (other spikes) during

immobility. Each dot represents the averaged modulation of a cell. The modulation strengths of

spikes participating in synchrony are significantly higher than those of spikes not participating in

synchrony (modulation strength for spikes participating in synchronous ensembles: 0.81±0.08, for

spikes not participating in synchronous ensembles: 0.63±0.12, mean±s.d., n=71 cells, p<0.001, pair

Student's *t*-test).

(B) Same as in (A), but during locomotion (modulation strength for spikes participating in

synchronous ensembles: 0.76±0.18, for spikes not participating in synchronous ensembles: 0.60±

0.12, mean±s.d., n=71 cells, p<0.001, pair Student's *t*-test).

(C) Average fluorescence triggered by cells' spikes participating in synchronous ensembles (red, sync

spikes) and not participating in synchronous ensembles during immobility (black, other spikes; n=59

cells). Inset: Pairwise comparison of average fluorescence 50±5 ms before time of spikes (t=0)

between the two triggered groups (n=59 cells).

(D) Same as in (C), but during locomotion.

Figure 5-figure supplement 1 Synchronization strengths during immobility show little correlation

with the similarity of spatial tunings between place cells. Synchronization strengths during

immobility plotted against correlation coefficients of spatial tunings for pairs of place cells.

Synchronization strengths show little correlation with similarities of spatial tunings (Spearman

correlation coefficient=-0.003, p=0.98, n=88 cell pairs).