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3 *In Vivo* Multi-Day Calcium Imaging of CA1 Hippocampus in Freely
4 Moving Rats Reveals a High Preponderance of Place Cells with
5 Consistent Place Fields
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Abstract: Calcium imaging using GCaMP indicators and miniature microscopes has

been used to image cellular populations during long timescales and in different task

phases, as well as to determine neuronal circuit topology and organization. Because the

hippocampus (HPC) is essential for tasks of memory, spatial navigation, and learning,

calcium imaging of large populations of HPC neurons can provide new insight on cell

changes over time during these tasks. All reported HPC in vivo calcium imaging

experiments have been done in mouse. However, rats have many behavioral and

physiological experimental advantages over mice. In this paper, we present the first (to

our knowledge) in vivo calcium imaging from CA1 hippocampus in freely moving male

rats. Using the UCLA Miniscope, we demonstrate that, in rat, hundreds of cells can be

visualized and held across weeks. We show that calcium events in these cells are

highly correlated with periods of movement, with few calcium events occurring during

periods without movement. We additionally show that an extremely large percent of

cells recorded during a navigational task are place cells (77.3±5.0%, surpassing the

percent seen during mouse calcium imaging), and that these cells enable accurate

decoding of animal position and can be held over days with consistent place fields in a

consistent spatial map. A detailed protocol is included, and implications of these

advancements on *in vivo* imaging and place field literature are discussed.

Significance statement:

In vivo calcium imaging in freely moving animals allows the visualization of cellular

activity across days. In this paper, we present the first in vivo Ca2+ recording from CA1

hippocampus in freely moving rats. We demonstrate that hundreds of cells can be

visualized and held across weeks, and that calcium activity corresponds to periods of

movement. We show that a high percentage (77.3±5.0%) of imaged cells are place

cells, and that these place cells enable accurate decoding and can be held stably over

days with little change in field location. Because the hippocampus is essential for many

tasks involving memory, navigation, and learning, imaging of large populations of HPC

neurons can shed new insight on cellular activity changes and organization.

Introduction

The explosion in the use of *in vivo* calcium imaging has allowed the visualization of hundreds of identified neurons over long time periods (Grienberger and Konnerth, 2012a; Grienberger and Konnerth, 2012b; Rickgauer et al., 2014; Hamel et al., 2015; Resendez et al., 2016; Aharoni and Hoogland, 2019b; Aharoni et al., 2019; Zhou et al., 2019). Calcium (Ca2+) imaging using GCaMP calcium indicators and miniature microscopes has been used to image cellular populations during long timescales (Dombeck et al., 2010; Ziv et al., 2013; Cai et al., 2016; Sheintuch et al., 2017; Hainmueller and Bartos, 2018; Kinsky et al., 2020; Sheintuch et al., 2020) and in different task phases (Zhang and Li, 2018; Jimenez et al., 2020; Shuman et al., 2020), as well as to determine neuronal circuit topography and organization (Klaus et al., 2017; Wang et al., 2017; Guo et al., 2020; Shin et al., 2020).

The hippocampus (HPC) has been studied with cellular resolution using *in vivo* electrophysiology, but this technique does not allow the same cells to be definitively followed over days, nor does it allow the visualization of cellular organization within the structure (O'Keefe, 1976; Wilson and McNaughton, 1993; McEchron and Disterhoft, 1999; Leutgeb et al., 2004; Schimanski et al., 2013). *In vivo* calcium imaging is particularly well suited for imaging the rodent HPC, as the orientation of the horizontal cell layer permits imaging of large numbers of neurons with insertion of a 1mm diameter or smaller lens (Guo et al., 2020; Kinsky et al., 2020; Sheintuch et al., 2020; Stefanini et al., 2020a). Because the hippocampus is essential for many tasks involving memory (Scoville and Milner, 1957; Squire, 1992; Eichenbaum et al., 1999; Hasselmo and McClelland, 1999; Ferbinteanu and Shapiro, 2003; Ferbinteanu et al., 2006; Smith and Mizumori, 2006; Vann, 2013; Cai et al., 2016; Josselyn and Tonegawa, 2020), spatial navigation (Foster et al., 2000; Rosenzweig et al., 2003; McNaughton et al., 2006; Foster and Knierim, 2012; Moser et al., 2017; Avigan et al., 2020), and learning (Moyer et al., 1990; Chan et al., 2001; Ito et al., 2005; Andrzejewski and Ryals, 2016), Ca2+ imaging of large populations of HPC neurons can shed new insight on cell changes and organization over time during these tasks.

To our knowledge, all reports of hippocampal *in vivo* calcium imaging have been

done in mouse (although calcium imaging has been used to visualize other structures in

rat (Scott et al., 2018; Anner et al., 2020; Hart et al., 2020; Pritchard et al., 2021)).

Using calcium imaging in rat hippocampus requires optimizing a number of

methodological parameters, including aspiration and lens implantation techniques, to

account for the increased thickness of the rat alveus compared to mouse which makes

focusing on the CA1 cell layer difficult (Swanson, 2004; Routh et al., 2009; Franklin and

Paxinos, 2013). However, rats have many basic experimental advantages over mice:

First, rats are able to learn and perform more complex tasks than mice, and employ

more advanced strategies, and are thus suited for a wider variety of behavioral

experiments (Whishaw, 1995; Frick et al., 2000; Whishaw et al., 2001; Cressant et al.,

2007; Rosenfeld and Ferguson, 2014; Hok et al., 2016). Second, compared to mouse

physiology, rat physiology for a number of conditions and disorders (including addiction,

aging, and schizophrenia, among others) more closely mirrors human physiology

(Iannaccone and Jacob, 2009; Vengeliene et al., 2014; Ellenbroek and Youn, 2016;

Carter et al., 2020). Third, rats are less stressed by handling than mice, thus introducing

less variability in experimental procedures (Buerge and Weiss, 2004).

Rats are also an advantageous organism to use for calcium imaging: Due to their

larger size, rats are able to support larger implants, including a wider diameter imaging

lens, larger and heavier cameras that may be available in the future, and more

combinations with electrophysiology implants. This advantage will thereby enable the

recording of a greater number of cells (Voigts et al., 2013; Aharoni and Hoogland,

2019b; Voigts et al., 2020). Additionally, rat hippocampal neurons are less excitable

than those in mice (Routh et al., 2009; Hok et al., 2016), and rats are less prone to post-

operative seizures than mice (Hunt et al., 2009, 2010; Bolkvadze and Pitkanen, 2012).

Rats may therefore, compared to mice, be at a decreased risk of the seizures (and

potentially to cell death) seen in mice after prolonged GCaMP expression (Tian et al.,

2009b; Grienberger and Konnerth, 2012b; Resendez et al., 2016; Steinmetz et al.,

2017; Yang et al., 2018).

Although *in vivo* Ca2+ imaging would appear to hold substantial promise for the

study of hippocampus in freely moving rats, there is currently no direct evidence that it

will work in this system. To our knowledge, there are no published studies employing *in*

vivo Ca2+ imaging in rat hippocampus, likely partially due to the number of

methodological changes necessary to adapt the technique from mice to rats. In this

paper, we present reports of *in vivo* Ca2+ recording from CA1 hippocampus in freely

moving rats. We demonstrate that hundreds of cells are reliably visualized and held

across weeks using this technique, and that calcium activity corresponds to periods of

movement. We show that during a navigational task involving retrieving food reward on

a linear track, a uniquely high percentage (77.3±5.0%) of imaged cells are place cells

(as compared to calcium imaging studies in mouse and electrophysiology studies in rat

and mouse), and that these place cells enable accurate decoding and can be held

stably over days with little change in field location. A detailed protocol for this technique,

including notes on the numerous parameter changes needed to use Ca2+ in rats, is

included in the Materials and Methods section, and implications of these advancements

are discussed.

Materials and Methods

SUBJECT DETAILS

All procedures were performed within Northwestern Institutional Animal Care and

Use Committee and NIH guidelines. Five male Fischer 344 x Brown Norway rats (275g

to 325g) were sourced from National Institute on Aging colony at Charles River

Laboratories, injected with AAV9-GCaMP7c, implanted with a 2mm GRIN lens and run

on a linear maze (Figure 1). Animals were individually housed in an animal facility with a

12h light dark cycle.

METHOD DETAILS

GCaMP7c Injection and Lens Implantation

Rats were anesthetized with isoflurane (induction 4%, maintenance 1-2%).

Following anesthesia, ear bars, and skull leveling, the skull was cleaned and a

craniotomy site was marked above CA1 (stereotaxic coordinates Bregma AP -4.00mm,

ML 3.00mm). The skull was scored around the craniotomy site using a razor blade to

provide texture for future application of epoxy. A craniotomy was made with a 2mm

trephine around the marked site and dura was removed with a bent 30 gauge needle. A

Hamilton syringe (Gastight® 1700 Series Syringes, Hamilton, Model 1705 Small RN

Syringe, Volume=50 μL, Point Style=3, Gauge=22s, Needle Length=50.8 mm) loaded

with AAV9-GCaMP7c (obtained from Viogene Biosciences, packaged AAV9 of pGP-

AAV-syn-jGCaMP7c-WPRE, titer $1.73*10^{13}$GC/mL) was lowered straight down to about

0.2mm past the CA1 cell layer (approximate coordinates Bregma AP -4.00mm, ML

3mm, DV 2.95mm relative to skull). 0.6uL of virus were injected over 12 minutes using

an automated stereotaxic Injector. The syringe was then raised 0.2mm (Bregma AP -

4.00mm, ML 3mm, DV 2.75mm) and another 0.6uL of virus were injected over 12

minutes (Figure 1a). The syringe was then left for 10 minutes before slowly raising to

the skull.

We used a computerized stereotaxic injection system that allowed us to specify

the coordinates of the injection. Due to the imprecision of craniotomies and imperfectly

level skulls, often the initial injection at the specified sub-cortical coordinates did not fall

in the center of the craniotomy. To increase and maximize the spread of viral injection,

we completed a second injection. The process of lowering the syringe and doing two

injections was repeated so that the two syringe "lowering points" were evenly spread

throughout the craniotomy site. There were thus a total of four injections of 0.6uL each

(2.4uL total): two injections per each brain entrance site, each at different depths

(Figure 1a).

Following injection, four skull screws were screwed into the skull: one rostral to

bregma above the same hemisphere as the craniotomy, two between bregma and

lambda on the hemisphere opposite the craniotomy, and one behind lambda on the

hemisphere with the craniotomy. Under a dissecting microscope, tissue from the

craniotomy was then aspirated using a vacuum pump and 25 gauge needle that had

been previously blunted using a Dremel drill. Tissue was quickly aspirated until the start

of horizontal striations could be seen in the craniotomy hole. Care was then taken to

"level" the hole and make sure aspiration was at an even depth. The horizontal

striations of the corpus callosum were then carefully and slowly aspirated until the

vertical striations of the alveus were visible in the entirety of the craniotomy hole (Figure

1b). If the aspiration hole was bleeding, a small twisted piece of gel foam was inserted

into the hole for approximately five minutes and then removed.

Notes differentiating this method from mouse Ca2+ imaging: We found the best

AAV for expression in rats was AAV9. Because of rat's lower seizure threshold and,

anecdotally noted, lower GCaMP expression than in mice, four injection sites appear

both necessary and sufficient for ideal GCaMP expression. The injection volumes are

much larger than those used in mice (Ziv et al., 2013; Cai et al., 2016), which is made

possible due to the larger hippocampal area and the technique of injecting below the

CA1 cell layer, instead of on it. Unlike in mice, where just the cortex or most dorsal part

of the corpus callosum is aspirated (Cai et al., 2016; Schoenfeld et al., 2021), it is

important that the entire corpus callosum be completely aspirated in order to get very

close to the alveus.

A 2mm GRIN lens (obtained from Go!Foton, CLH lens, 2.00mm diameter, 0.448

pitch, working distance 0.30mm, 550nm wavelength) was then held on the end of a

pipette using suction from the vacuum pump. The pipette was placed in the stereotaxic

instrument which was zeroed with the bottom of the lens on bregma. The lens was then

moved to the center of the craniotomy and lowered to a depth of approximately 3mm

from skull surface. The lens was then fixed in place with UV curing epoxy and the

vacuum pump was turned off and the pipette holding the lens removed. The epoxied

lens was then secured to the skull screws using dental acrylic. Animals were then given

buprenorphine (0.05mg/kg) and 20mL saline, taken off anesthesia, and allowed to

recover in a clean cage placed upon a heat pad.

Notes differentiating this method from mouse Ca2+ imaging: A much larger lens

(2mm diameter as opposed to 0.5mm or 1.0mm diameter) lens can be used in rats due

to their larger size. Of high importance: in order to be as close to the CA1 layer as

possible, the lens is inserted further down (about .5mm) lower than the cell layer. This

serves, presumably, to temporarily compress the alveus to make sure the alveus is

pressed against the grin lens; this allows for in focus imaging of the CA1. We saw no

permanent compression of the CA1 cell layer (Figure 1C) or any behavioral deficits that

are associated with HPC compression (Shim et al., 2003; Chen et al., 2017). We had

zero visualization of cells unless the lens was deeply inserted. Additionally, unlike in

mouse implantation surgeries, we did not have success using meloxicam after surgery

as it appeared to cause excessive bleeding around the lens.

Four to six weeks after surgery, animals were again anesthetized with isoflurane

and checked for cells expressing GCaMP. If expression was seen, the animal's lens

was base-plated. The baseplate was attached using UV-curing epoxy and dental

acrylic.

Notes differentiating this method from mouse Ca2+ imaging: During base plating,

it gets progressively harder to see cells the longer the animal is under anesthesia.

Additionally, while a few cells will appear to be bright circles in the captured image, cells

do not appear to be flashing at all while the rat is under isoflurane, even after a tail pinch

or loud noise (techniques often used to incite a calcium burst during mouse base

plating). We recommend focusing on the bright, not flashing cells while base plating.

The entire field of view should become progressively brighter as the animal wakes up.

We recommend not imaging for experiments within 24 hours of isoflurane exposure and
base plating.

Behavioral Training

Prior to training, animals were food deprived to 85% body weight and maintained
at this weight throughout the experiment. Implanted animals were trained on a 2.64m
Med Associates linear track with pellet dispensers at either end. Animals learned to run
back and forth between the sides of the track and received a grain pellet at alternate
ends. Each pellet received was considered a trial. The fasted animals were able to
achieve over 120 trials in 30 minutes. (Animal were fed after daily training to maintain
weight.) Six sessions were selected for analysis for each animal, with first and last
recorded sessions automatically included. When not being run, animals were housed in
individual cages with a 12hr light- 12hr dark light cycle.

Six sessions from each animal were selection for analysis. Although each animal
was run for at least seven sessions, occasionally an animal would refuse to run or run
slowly (often the day following a run with a lot of pellets received). For the animal run
the least number of times, only six sessions exceeded 50 trials per run. For consistency
across analysis, we therefore selected the six sessions with the most trials to analyze
from each animal. On average, the four animals ran 100.3+-33.1 trials per session, with
the individual animals averaging 94.7+-35.3, 85.5+24.0, 114.5+-19.8, and 106.2+-41.2
trials per session.

Because the runs with the most trials were selected for analysis, trials were not
always consecutive. If the first analyzed session is considered day 1, the analyzed
sessions were as follows for each animal (as numbered by days on a calendar; animals
were not necessarily run every day due to scheduling and equipment troubleshooting):
Animal 1:   Days 1, 2, 3, 7, 8, 9 *(run across 9 days)*
Animal 2:   Days 1, 2, 6, 7, 9, 13 *(run across 13 days)*
Animal 3:   Days: 1, 2, 11, 14, 15, 16 *(run across 16 days)*
Animal 4:   Days: 1, 2, 10, 14, 15, 16 *(run across 16 days)*

Calcium Imaging

One photon calcium imaging was done using UCLA V4 Miniscopes (Silva, 2017;

Aharoni and Hoogland, 2019a), assembled with standard lens configurations (two 3mm

diameter, 6mm FL achromat lens used in the objective module and one 4 mm diameter,

10mm FL achromat lens used in the emission module). We tested Ca2+ recording when

one of the objective module lenses was replaced by a 3mm diameter, 9mm FL

achromat lens, but no improvement in cell visualization was noted. Gain was set at high

and Miniscope LED intensity varied between 10 and 38.

Notes differentiating this method from mouse Ca2+ imaging: Base fluorescence

appears lower in rats as compared to mice. While we have been advised that, using

Miniscopes, gain can often be on "low" for mouse imaging, we found it necessary to

have gain on "high" to visualize the highest number of cells.

GCaMP Visualization in Slice and Immunohistochemical Processing of Tissue

Sections

Following conclusion of the study, animals were anesthetized and perfused with

PBS then 10% formalin. The brain was then sliced into 60 micron coronal sections. For

visualization as seen in Figure 1c, sections were mounted, let dry, and coverslipped

with Vectashield. Sections were then imaged with a confocal microscope.

For GCaMP and PV+ visualization in Figure 1d, all solutions were prepared in

phosphate buffered saline (PBS) containing 0.05% sodium azide and 2% normal

donkey serum. Free-floating sections were rinsed in PBS three times for a total of at

least 30 min between incubation steps. Tissue sections were sequentially placed in (1)

0.2% Triton X-100, 90 min; (2) mouse monoclonal anti-parvalbumin 2000X (Millipore),

48 hours; (3) rabbit anti-GFP 2000X,(AbCam), 24 hours; (4) biotinylated goat anti-rabbit

serum 200X (Jackson Immunochemicals), 90 min; (5) a cocktail containing Cy3

conjugated donkey anti-mouse serum 200x (Jackson Immunochemicals) and FITC

conjugated avidin 200x (Jackson), 90 min. Sections were coverslipped with a 9:1

mixture of glycerine and PBS containing 0.2% p-phenylenediamine and examined with

a confocal microscope.

QUANTIFICATION AND STATISTICAL ANALYSIS

Means are expressed as mean ± the standard deviation. Overall averages are computed by finding average per animal then averaging those values. All analysis code is available at https://github.com/hsw28/ca_imaging.

Position and speed analysis

Position was sampled by an overhead camera at 30Hz. Position tracking was done post-recording using Bonsai (Lopes et al., 2015). Speed was determined by taking the hypotenuse of the coordinates of the point one before and one after the time point of interest. Speed was then smoothed using a Gaussian kernel of 1s and speed was converted from pixels/s to cm/s.

Video pre-processing

Videos were recorded with Miniscope software at 15 frames/second and 752 by 480 pixels. All video processing was done using the Matlab-based CIATAH software (Ahanonu, 2018; Corder et al., 2019) and processing files are available at https://github.com/hsw28/ca_imaging/ . Videos were down sampled in space (3x using bi-linear interpolation) and time (2x). The video was normalized by subtracting the mean value of each frame from the frame. Each frame was then spatially filtered (normalized) using a bandpass FFT filter (70-100cycles/pixel). The video was then motion corrected to a chosen reference frame using TurboReg (Thevenaz et al., 1998). Videos were then converted to relative florescence (dF/F_0), where F_0 was the mean over the entire video.

Cell identification

Cells were first automatically identified using the Matlab-based CIATAH software (Ahanonu, 2018; Corder et al., 2019), using the CNMF-E method (Zhou et al., 2018).

Processing files are available at https://github.com/hsw28/ca_imaging/ . Images were

filtered with a gaussian kernel of width 3 pixels. Neuron diameter was set at a pixel size

of 13. The threshold for merging neurons was set at a calcium trace correlation of 0.65,

and neurons were merged if their distances were smaller than 2 pixels and they had

highly correlated spatial shapes (correlation>0.8) and small temporal correlations

(correlation <0.4).

All cells identified using CNMF-E were then scored as neurons or not by a

human scorer. Scoring was also done within the Matlab-based CIATAH software

(Ahanonu, 2018; Corder et al., 2019) in a Matlab GUI. Human scoring was done while

visualizing an activity trace, average Ca2+ waveform, a cropped movie montage of the

candidate cell's Ca2+ events, and a maximum projection of all cells on which the

candidate cell was highlighted. The relative fluorescence ($\Delta F/F_0$) local maxima of each

identified cell were considered calcium event times.

Place cell identification and spiking characteristics

Place cells were identified using two criteria. First, cells must have a mean firing

rate greater than 0.01Hz during periods of movement (>12cm/s) (Wirtshafter and

Wilson, 2019, 2020). Second, the mutual information (MI) computed in either direction

of travel must be greater than 95% of MI scores computed 500 times from shuffled cells

(Kinsky et al., 2018). To compute MI, the track was divided lengthwise into 4cm bins,

the firing rate of each cell and the occupancy of the animal were found for each bin.

Rate and occupancy were smoothed with a 2cm Gaussian kernel. Mutual information

was computed during periods of movement in each direction of travel as follows

(Olypher et al., 2003; Kinsky et al., 2018; Wirtshafter and Wilson, 2020):

$P_s = calcium\ event\ probability\ in\ each\ bin$

$P_o = occupancy\ probability\ at\ each\ bin$

$p = \dfrac{P_s}{P_o}$

$M_s = \sum P_s$

$$M_o = \sum P_o$$

$$MI = \sum p * \log_2\left(\frac{p}{M_s * M_o}\right)$$

Place field location was determined by the location of maximum firing rate after

binning position into 4cm bins. Speed modulation for individual cells was determined by

computing average event rate when the animal was moving <=5cm/s and >=12cm/s; a

cell was considered speed modulated if the rate at >=12cm/s was larger. At the

population level, speed modulation was determined by examining population event rate

versus speed. Cross correlations between speed and event rate were computed by

binning speed and rate into 100ms bins.

Position decoding

Position was decoded using all recorded cells in a session and was decoded for

times when the rat was moving (velocity>12cm/s). The track was split in 10cm

segments and the time window of decoding was 0.5 seconds with non-overlapping time

bins. Decoding was done using Bayesian decoding (Zhang et al., 1998).

$x = positions, f_i(x) = average\ firing\ rate\ per\ second\ for\ cell\ N\ at\ position\ X$

$n = calcium\ events\ per\ cell, n_i = number\ of\ events\ in\ time\ window$

$N = total\ number\ of\ cells$

$t = time\ window\ of\ decoding$

$C(t, n) = normalization\ factor\ for\ probabilities\ to\ equal\ 1$

$P(x|n) = odds\ of\ being\ at\ position\ x\ giving\ calcium\ events\ n$

$$P(x|n) = C(t,n)\left(\prod_{i=1}^{N} f_i(x)^{n_i}\right)\exp\left(-t\sum_{i=1}^{N} f_i(x)\right)$$

Accuracy decoding was determined by randomly shuffling firing per position bins

and re-decoding. The difference between actual and decoded position was then

determined per time bin and compared to decoded shuffled data.

Validation across days and within session

Validation was done using the Matlab-based CIATAH software (Ahanonu, 2018; Corder et al., 2019). Videos underwent 3 rounds of registration comprised of Turboreg image rotation (Thevenaz et al., 1998). An image binarization threshold of 40% of the images' maximum value was used to remove background noise and axons and dendrites. A distance threshold of a maximum of 5 pixels was used to match cells across sessions, with a minimum 2-D correlation coefficient of 0.5. All sessions were aligned to the fourth recording session.

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Results

To initiate Ca2+ imaging, we injected AAV9-Syn-GCaMP7c below the CA1 cell

layer in five male Fischer 344 x Brown Norway rats (Figure 1a, see Materials and

Methods) (Nathanson et al., 2009). After injection, in the same surgery, cortex and

corpus collosum were aspirated through a craniotomy until the alveus was clearly visible

(Figure 1b, see Materials and Methods). A 2mm diameter GRIN lens was inserted into

the craniotomy and cemented in place. Three animals were checked for GCaMP

expression four weeks after surgery, and all three animals showed expression and were

subsequently affixed with a base plate to secure the position of the Miniscope. The

other two animals were checked six weeks after surgery and showed expression and

were base plated. After the completion of the experiment, animals were sacrificed and

expression of GCaMP in CA1 pyramidal cells was verified with microscopy (Figure 1c).

Although contributions of interneurons cannot be ruled out, antibody staining for

parvalbumin positive (PV+) interneurons revealed that these cells constituted only a

small proportion of GCaMP cells in the stratum pyramidale (Figure 1d), consistent with

observations that <5% of cells in the pyramidal layer are interneurons (Aika et al.,

1994). GCaMP expression also appeared to be primarily limited to pyramidal cells in the

cell layer.

Animals were then food deprived to 85% body weight and trained on a 2.64m

linear track (Figure 1e). Animals were rewarded with grain pellets on alternating ends of

the linear track and quickly learned to run back and forth on the track. A visit to one

reward well signified a trial. Animals were run for a minimum of 30 minutes or 50 trials,

whichever came later, with total track time not to exceed one hour per session. Cells

were recorded for 7 days for one animal, 11 days for one animal, and 16 days for two

animals (n=4 rats, one of the five animals was eliminated when the implant was lost).

The six sessions with the most trials run were analyzed from each animal

(average trials run = 100.3±33.1, see Materials and Methods), with first and last

recorded sessions automatically included; sessions were not always consecutive (see

Materials and Methods). These sessions were analyzed for cellular activity using the

CNMF-E method implemented by the CIATAH software package (Ahanonu, 2018; Zhou

et al., 2018; Corder et al., 2019) (See Materials and Methods). A total of 5761 neurons

were recorded over the 24 sessions, with a maximum of 428 cells from a single session

(mean of 239.6±90.0 cells) (Figure 2a-d). The number of cells recorded from each

animal across sessions was relatively constant (Figure 2c).

We then determined average calcium events/s across animals to be 0.090+-

0.019 events/s, with individual animals averaging 0.084+-0.024, 0.074+-0.026, 0.085+-

0.020, and 0.116+-0.032 events/second. (Figure 2D-E). Some difference in average

calcium event rate could be accounted for by differences in the base line speeds of the

animals: there was a significant positive correlation between the animal's average

speed and the average events/s for each session ($F_{(22)}=4.2$, $p=0.05$) (Figure 2F).

Because speed is highly variable on a moment-to-moment basis, we also completed a

cross-correlation between event rate and animal speed. In every session, we saw a

positive cross-correlation between the animal's speed and the number of calcium

events across the population. The average cross-correlation at a lag of 0s was 1.8+-

0.08, although the average maximum cross-correlation was 0.2 +-0.7 at an average lag

of +0.47+-0.54s, indicating that calcium events reliably followed changes in speed, as

opposed to vice versa (Figure 2G).

Significantly, calcium activity appeared to be heavily modulated by the animal's

activity, with periods of higher activity when the animal was running rather than

stationary on the track (Figure 3). The biggest contributor to a change in event rate was

transition from no movement to movement, with very few events occurring during

periods of no movement (Figure 3C).

We then determined if there were place cells on any of the trials. A cell was

considered to be a place cell if the mutual information (MI) computed in either direction

of travel was greater than 95% of MI scores computed 500 times after shuffling the

times of the calcium events that occur when the animal is moving (Ziv et al., 2013;

Kinsky et al., 2018). Based on this criterion all trials contained place cells (Figure 3a).

An average of 77.3±5.0%, of all cells recorded on the track were place cells (Figure 3),

with individual rat averages at 71.6±10.6%, 77.3±1.2%, 76.4±9.5%, and 83.7±7.1%. We

additionally determined the percentage of place cells if we only analyzed cells that had

an average calcium events rate of greater than 0.01hz (Stefanini et al., 2020a). This

additional criterion yielded an even higher percentage of place cells, with 88.3±4.4% of

these cells being classified as place cells. (Figure 3a).

We extended these results using an additional criterion previously used to

determine the percentage of place cells using calcium imaging (Stefanini et al., 2020a).

We determined the percentage of cells that would be place cells if the mutual

information value exceeded the mean plus three standard deviations of a shuffled

distribution. Using this criterion, we found that 64.3+-5.1 percent of all cells were place

cells, with individual animals at 59.5+-12.0%, 65.5+-2.3%, 61.2+-14%, and 71.0+-

12.1%. If we applied the 0.01hz threshold for cell inclusion to this group, 73.4+-4.3% of

cells were place cells, with individual animals at 68.0+-11.8%, 77.6+-5.6%, 77.0+-14.6%

and 76.0+-13.0%.

We wished to further investigate the properties of these place cells and to

confirm that cells were increasing their firing rate in place fields as is typical for

pyramidal cells; as opposed to interneurons that mark spatial location by a substantial

decrease in firing (Ego-Stengel and Wilson, 2007; Wilent and Nitz, 2007; Dombeck et

al., 2010)). We examined the place cells' maximum event rate during movement

(presumably occurring in the place field) with their average firing rate during movement.

The mean maximum rate was 0.15events/s, and the mean average rate was

0.02events/s. On average, cells increased firing 7.4x from their average firing rate

during movement to their maximum firing rate (Figure 4D). We next determined if speed

modulation occurred in the cells which display place fields. We compared whether

speed-modulated neurons increased their in-field firing rate as much as non-speed-

modulated neurons. We found no significant difference in firing rate increases between

speed-modulated and non-modulated neurons (t-test (4023)=0.7102, p>0.05), nor were

the distributions of increases different (KS-test p>0.05): both speed-modulated and non-

modulated cells increased their firing rates, on average, >7x from baseline at the center

of their place fields.

We then aimed to determine whether place cell firing was sufficient to decode the

position of the animal. We split the track in 10cm segments and used Bayesian

decoding to determine the position of the animal in non-overlapping 0.5 second

increments (Zhang et al., 1998). We found that the best positional decoding was

accurate to a median error of 4.3cm (1.6% of the 2.64m track). Across all animals,

accuracy ranged from a median of 4.3cm to 13.6cm (decoding errors were below 10cm

in 23 sessions and above 10cm in one session) with a median error of 5.9cm (Figure

5a). To confirm that decoding accuracy was better than chance we shuffled firing rate

per position bins and, using the shuffled data, decoded position. Median error for

shuffled data ranged from 59.1cm to 77.6cm, which was significantly worse than non-

shuffled data (Kruskal-Wallis test, $x^2(47)= 35.3$, $p=2.9*10^{-9}$) (Figure 5b).

We then aimed to determine if the same cells could be held over the duration of

our recordings (six recording sessions, spanning 7, 11, or 16 days). 40% of the cells

imaged on day one of the 7-day session were also present on day seven. Of the rat run

for 11 days, 31% of the cells present on day one were also present on day eleven. For

the two rats run 16 days, 28% and 50% of cells, respectively, that were present on day

one were also present on day sixteen. Across all four animals, cells that were imaged

on day one of recording had, on average, a 72±6% chance of being imaged on at least

one subsequent day of recording (Figure 6A). (Of note, while the camera was placed in

the same approximate position each day, thus allowing landmarks (such as blood

vessels) to be followed throughout the study, we did not attempt to perfectly align

landmarks or focal planes across days; instead recordings each day were made to

optimize total cell number in each day as proof of method.)

We then looked to determine how often a cell appeared in each session. Across

all cells, 59.4+-5.4% of cells were visible in only one session, 19.9+-2.4 were visible in

two, 11.2+-1.5 in three, 6.3+-0.8 in four, 2.3+-1.3 in five, and 1.0+-0.9 were in all six

sessions (Figure 6B). We then looked at place cells that were repeated across

sessions. As stated, 22.6+-4.3% of cells did not have place cells in any session. Of the

77.4% of cells that did have place fields, 61.0+-5.7% (47.2% of total cells) had fields in

only one session, 20.6+-2.0 (15.9% of total) had fields in two sessions, 10.8+-2.4%

(8.3% of total) had fields in three sessions, 5.1+-0.7% (4.9% of total) had fields in four

sessions, 2.0+-1.4% (1.5% of total) had fields in five sessions, and 0.5+-0.5 (0.4%of

total) had fields in all six sessions (Figure 6C). Place cell recurrence was not statistically

different from the recurrence of all cells (paired t-test(5)=2.04*10^{-15}, p=1).

We then determined the recurrence probability of a cell based on the number of

days apart the recording took place (Figure 6D). Three of our animals had relatively

consistent recurrence probabilities, while one animal (animal 3) was highly variable.

Going back to the recordings, we saw that recordings from this animal came from two

different focal planes, causing the recording of two different cell populations. Excluding

this animal, there was a modest trend across days for a decreasing recurrence

probability, with the probability of a cell reoccurring decreasing by 0.43% per day

separating the recording periods (Figure 6E).

We then determined whether a cell's place field remained in the same location

during days when the place cell reoccurred. We found that, across all days of recording,

the mean standard deviation of place field locations was about 36cm (~13% of track

length), with the individual animals at 39cm, 36cm, 32cm, and 37cm. We compared

these distances with distances if place field locations were shuffled, and we repeated

shuffling 500 times. All four animals had a smaller deviation between fields than would

be expected by chance (p<0.05 for all animals). Between individual days (including

nonadjacent days), the change in field distance was highly variable at 65cm+-71cm,

with a median of 40cm and a large concentration of values around 0cm (Figure 6F).

Although fields across days were more consistent than would have been

expected by random chance, we had not yet determined whether the animal had a

consistent map across all recording days (Ziv et al., 2013), or if it switched between

multiple maps (Sheintuch et al., 2020). To determine this, we compared place field

location for cells in a session against the place field locations of the same cells in

another session. To do this, for each cell, we plotted the location of the place field for

that session against the place field location of the same cell for another session. We did

this comparison for every cell that had a place field in more than one session. If place

field mapping was similar from session to session, we would expect the plotted field

locations of one session to have a strong positive correlation with the locations from

another session. Of the pairs of sessions that had more than 10 place cells shared

527 between sessions (57 out of 60 total session pairs—15 session pairs for rat 1, 15 for rat

528 2, 14 for rat 3, and 13 for rat 4), 53 of these pairs (93%) had a significant positive

529 correlation (f-test, $p<0.05$) between the place field locations, with no evidence of

530 switching between maps across days (Figure 6G). We found that correlations between

531 field locations did not decay as a function of the number of days between sessions

532 (Figure 6H).

535 **Discussion**

537  In this manuscript, we have demonstrated calcium imaging recordings can be

538 performed in the CA1 region of the hippocampus in freely moving rats. These

539 recordings can capture hundreds of cells in a single recording session, including a high

540 number and percentage of place cells sufficient to decode the animal's position to within

541 centimeters of accuracy, and that place cell field location is consistent across days.

542 Calcium imaging in rat will allow for more behaviorally complex and translational

543 experiments than have been done in mouse.

545 **Measuring Spatial Content**

546  We chose to use mutual information as the measure of spatial content for

547 multiple reasons: First, it has precedence for use in both calcium imaging (Ziv et al.,

548 2013; Meshulam et al., 2017; Kinsky et al., 2018; Kinsky et al., 2020; Stefanini et al.,

549 2020a) and electrophysiology studies (Jun et al., 2020) (Table 1). Second, threshold

550 measures commonly used in electrophysiology that depend on firing rate (such as "a

551 cell is a place cell if firing rate, as measured by electrophysiology, exceeds 2Hz for a

552 10cm area") (Rich et al., 2014) (Table 1) do not transfer well to calcium imaging

553 experiments as, using currently availably GCaMPs, calcium events are captured much

554 less frequently than spikes in electrophysiology (typically, ~0.02-0.06 events/s is the

555 mean rate observed during 1P calcium imaging, versus >0.5-1.5 spikes/s in

556 electrophysiology, a 25-fold difference, see Table 1). Third, other measures of

information content, such as bits/spike or bits/second (Table 1), tend to be highly

sensitive to changes in firing rate, including baseline firing rate (Souza et al., 2018), as

compared to mutual information. Specifically, cells with lower baseline firing rates are

favored to have higher baseline bits/spike or bits/second values (Souza et al., 2018;

Wirtshafter and Wilson, 2020), which makes it not a useful metric with which to compare

values from calcium imaging with data collected from electrophysiology. Additionally,

mutual information is advantageous in that it is a better predictor of decoding accuracy

than bits/spike and bits/second, and thus may better capture spatial informational

content (Souza et al., 2018). The criteria we used to identify place cells were as or more

stringent than previously used criteria in calcium imaging that identified a significantly

lower percentage of place fields (Table 1).

It is well established that pyramidal place cells increase their firing rate in their

place field, while interneurons express place specificity by decreasing firing rate. Given

that the majority of cells we observed displayed an increased rate of calcium events at

specific locations, it is reasonable to conclude that these were almost certainly

pyramidal cells in the stratum pyramidale. This conclusion is further supported by the

fact that many of the labeled cells displayed an obviously pyramidal morphology (Figure

1c), the infrequency of double labeling of GCaMP+ and PV+ cells (Figure 1d), and by

findings that only about 3% of cells in the stratum pyramidale are interneurons (Aika et

al., 1994). Additionally, GCaMP expression appears to be limited to pyramidal cells in

the cell layer, although it is unclear if this is a result of injection location or viral

specificity. As such, while it cannot be definitively determined if interneurons have

contributed to the observed effects, their numerical contribution would, at most, be very

small.

The percentage of place cells in rat calcium imaging exceeds the percentage in

mouse calcium imaging

Significantly, we found a higher percentage of cells were place cells in our

calcium imaging experiments in rats than those reported in mouse CA1 calcium imaging

experiments, even after the mice experiments were corrected for cells with low spiking

rates (Table 1). (The studies with mice reported 20-30% of imaged CA1 cells were

place cells, which is also substantially below most electrophysiology experiments (Ziv et

al., 2013; Stefanini et al., 2020b; Stefanini et al., 2020a)). Previous work has reported

that mouse place cells are less spatially tuned, contain less spatial information, and

have less stable place fields than in rat (Hok et al., 2016; Mou et al., 2018), so some of

the difference in place cell percent may be due to species differences in spatial

representations. However, there are also likely physiological differences between rats

and mice that affect the type of activity captured in calcium imaging recordings. CA1

cells in mice are more excitable than those in rat (Routh et al., 2009; Hok et al., 2016),

and prolonged GCaMP expression in mice has been shown to lead to cell

hyperexcitability, seizures, and even cell death (Tian et al., 2009b; Grienberger and

Konnerth, 2012b; Resendez et al., 2016; Steinmetz et al., 2017; Yang et al., 2018). The

higher baseline excitability of mouse cells, combined with a further increase in cellular

excitability after GCaMP expression may result in a lower threshold for cell firing and

less defined place fields, less stable fields, and, therefore, a seemingly lower number of

place cells compared to electrophysiology recordings (Hok et al., 2016). Because rats

have a lower baseline excitability, they may be less prone to these GCaMP effects,

although further slice electrophysiology work will need to be done to directly compare

the effect of GCaMP on mouse and rat neurons.

We considered that differences in calcium event rate between our data and other

studies might lead to the high percentage of place cells we saw in this work. On one

end of the spectrum, Stefanini et al 2020 found a particularly low percentage of place

cells (~4% of recorded cells in mouse CA1, using calcium imaging) and investigated

whether this might be a consequence of collecting many cells with low calcium event

rates. Their reported mean rate was 0.06+-0.04 events/s as recorded in mice in an open

field, consistent with 0.05-0.5 events/s in other calcium imaging experiments in mouse.

Even after eliminating cells with <0.1events/sec (a 10x less stringent criteria than we

used in Figure 3A), they still only found that 10-20% of cells were place cells (Stefanini

et al., 2020a). Conversely, Ziv 2013, who reported a high (for mouse) 20% of imaged

CA1 mouse cells were place cells, reported a higher mean event rate, at ~0.1-0.5

events/s (Ziv et al., 2013). Our results (0.090+-0.019 events/s, Figure 2D-E) are on the

lower end of those reported in Ziv et al, and potentially consistent with the fact that rat

neurons are somewhat less excitable than those in mice (Routh et al., 2009; Hok et al.,

2016). Additionally, our very slightly lower event rate could be the result of less 'out of

field' firing of place cells, which also tends to be more predominant in mice (Hok et al.,

2016). However, our results are within the range of those reported in mouse imaging

studies (Table 1), so differences in event rate are not a likely to be a major contributor to

the place cell percentage we observed.

In addition to potentially allowing the recording of a higher number of place cells,

the lower excitability of rat CA1 neurons, compared to mouse, also has other

advantages. We saw no behavioral evidence of post-operative seizures in any of our

implanted animals, and no behavioral or imaging evidence of seizures several months

post viral injection and lens implementation. This is in contrast to calcium imaging

experiments in mice, which frequently report that animals have to be removed from

behavior or analysis due to the development of seizure activity (Steinmetz et al., 2017;

Huang et al., 2021).

Differences in place cell findings between electrophysiology, intermediate early

gene studies, and calcium imaging

The percentage of hippocampal CA1 cells found to be place cells using calcium

recordings (77.3+-5.0%) was also very slightly above the upper range of findings from

electrophysiological data in rats, which report that 35-75% of CA1 pyramidal cells

recorded on a navigational task are place cells (the number varies widely depending on

task parameters, with studies using a linear track task reporting the highest number of

place cells) (Table 1) (Markus et al., 1994; Fenton et al., 2008; Wirtshafter and Wilson,

2020). There are multiple possible reasons related to recording technology as to why

our calcium imaging experiments detected a higher percentage of place cells than

traditional electrophysiology experiments: the difference likely resulted from

electrophysiology capturing (as compared to calcium imaging) both a lower number of

cells with active place fields detected (a smaller numerator) and a larger number of cells

without active place fields detected (a larger denominator). First, in electrophysiology

experiments with moveable electrodes (such as tetrodes), electrodes are typically

positioned while rats are inside a "sleep box" or other non-training apparatus until cells

are detected. The animal is then placed on a track for experimentation. This practice

optimizes cell detection for place cells that are active in the "sleep box" (both place cells

with fields in the sleep box, and place cells that are reactivated during ripples during

sleep), rather than cells that have place fields on the track. Adjusting and recording in a

sleep box therefore increases the number of total cells recorded that do not have fields

on the track and therefore reduce the place cell percentage. Because calcium imaging

uses a fixed lens location and images a large number of cells at once, recording is not

optimized for a particular environment. This may result in the recording of a higher

percentage of active place cells on the recording apparatus. (Due to logistical

constraints in this study, we did not complete any recording while the animal was not on

the track.) Second, while an effort is made in electrophysiology to discard times on the

track when the animal is not moving, and thus to remove corresponding spiking during

ripples, this removal is never perfectly done. The recording of ripples during

electrophysiology has two effects: the recording of cells that are being replayed on the

track but do not have place fields on the track and, second, the recording of 'extra-field'

ripple spiking from cells with fields on the track. This 'extra-field' spiking can cause

place fields to seem less defined, and cause corresponding distortion of measures of

spatial information, such as mutual information and bits/spike. This, in turn, may cause

these cells not to be classified as place cells. Conversely, in calcium imaging, very few

to no calcium events are picked up during periods of no movement (see later in

discussion, figure 3), so both of these issues (increasing the count of cells without fields

on the track, and the distortion of spatial information for place cells) are avoided. Third,

while current GCaMPs, such as the GCaMP7c used in this study, have improved single

spike resolution compared to past GCaMPs, they still fall short of the single spike

resolution of electrophysiology (Tian et al., 2009a; Chen et al., 2013; Podor et al.,

2015a; Podor et al., 2015b; Shemesh et al., 2020). This property of calcium indicators

causes individual spikes, such as those that may occur on the track outside a cell's

place field, to be missed, therefore biasing calcium imaging towards fast cell activity that

only occurs in the confines of a place field. Calcium imaging therefore, again, results in

potentially higher spatial information metrics as compared to electrophysiology, which

may cause the identification of more place cells. Finally, compared to electrophysiology,

calcium imaging provides additional characteristics (cell shape, size, and brightness) by

which to separate adjacent or nearby cells. Because conflation of multiple cells in

electrophysiology may disguise place fields and disrupt place field measurements (such

as mutual information or bits per spike), the additional resolution afforded by calcium

imaging may allow for more accurate place cell differentiation and identification.

Similar to electrophysiology and in contrast to intermediate early gene (IEG)

studies, calcium imaging is biased towards capturing cells active in an environment and

not the total cellular population. Additionally, as discussed above, due to limits on single

spike resolution, calcium imaging may be picking up even fewer cells with infrequent

spiking compared to electrophysiology. Thus, this bias of calcium imaging toward

detecting only active cells yields a higher percentage of place cells than intermediate

early gene studies that estimate a total of 30-40% of cells to be active in any given

environment (table 1) (Guzowski et al., 1999; Witharana et al., 2016).

1P calcium imaging using GCAMP7c is movement responsive and minimally

activated when the animal is still

Importantly, while calcium transients and events are correlated with action

potentials, they are far from a 1:1 correspondence and may capture more or less

information than electrophysiology (Huang et al., 2021). During our recording sessions,

we saw clear evidence that calcium event rate was correlated with periods of movement

(Figure 3). In electrophysiology experiments, place cell spiking is generally linearly

correlated with running speed (McNaughton et al., 1983; Huxter et al., 2003), and

significant spiking occurs during ripples when the animal is not moving (Ylinen et al.,

1995; Chrobak and Buzsaki, 1996; Csicsvari et al., 1999; Kudrimoti et al., 1999). Our

finding greatly differs from the patterns of spiking seen in electrophysiology data, as we

saw minimal calcium events during periods of no movement, with a pattern of

logarithmic growth of the number of calcium events as the animal began moving and

sped up (Figure 3C). Because the vast majority of CA1 pyramidal cells are speed and/or

ripple modulated, and we have imaged primarily pyramidal cells, this result is likely not

explainable by the population of cells that express GCaMP (Huxter et al., 2003). Our

result is consistent with calcium imaging experiments done in mice by Zhou et al., in

which more calcium events were seen during movement in an environment where food

reward could be obtained than during quiet wakefulness in a neutral box (Zhou et al.,

2019). In our experiment, periods of quiet wake were in the same environment (the

track) as periods of movement, which shows that changes in calcium event frequency

are related to movement and not to contextual associations. Zhou et al. posit that the

high amount of calcium activity that occurs during movement, but not during sharp wave

ripples in mouse, implicate calcium events in theta dependent processes, such as

memory encoding and planning (Buzsaki, 1989; Hasselmo et al., 2002; Hasselmo,

2005; Shirvalkar et al., 2010; Colgin, 2013; Wikenheiser and Redish, 2015; Zhou et al.,

2019; Wirtshafter and Wilson, 2021). It is possible, and consistent with slice data, that

NMDA receptors, which permit calcium influx, are minimally involved in cell spiking

during ripples, and the process is more heavily modulated by AMPA receptors, which do

not permit calcium influx (Maier et al., 2003; Colgin et al., 2004). There is also evidence

that activity during ripples can be visualized using 2-photon calcium imaging in mice

(Malvache et al., 2016; Grosmark et al., 2021), although we could not find any

experiments showing this activity using 1-photon imaging. It is therefore unclear

whether limits to imaging spiking activity during ripples is due to the imaging technique

or to the use of rats. Further work will be needed to determine if, in rat, frequent calcium

events are also unique to movement, as opposed to sharp wave ripples.

Results show stable place cell identification and these cells have stable place

fields over days

Our findings on cell activity over days were similar to and consistent with prior

results reported in mice. Prior work has reported that about 57% of cells are active in

only one or two sessions (Ziv et al., 2013); we found a higher proportion (59.4+-5.4% of

cells were visible in only one session, 19.9+-2.4 were visible in two) (Figure 6B), likely

because we worked to optimize cell count over precisely consistent focal planes. We

also reported that probability of a cell reoccurring decreased by 0.43% per day

separating recording sessions. This result was exactly consistent with prior results in

mouse that cell reoccurrence decreases from ~25% for sessions 5 days apart to ~15%

for sessions 30 days apart (25 days difference * 0.43 = ~10.8% change)(Figure 6E) (Ziv

et al., 2013). This potential cell turnover did not impact the accuracy of the

representation of the environment, as decoding accuracy was consistent throughout

sessions (Levy et al., 2021).

There are differing results regarding whether an environment has a single semi-

stable map (Ziv et al., 2013; Mau et al., 2018; Kinsky et al., 2020), or there are multiple

maps, involving the same cells, that code for a single environment (Sheintuch et al.,

2020). In this study, we only saw evidence of the former, as the vast majority of

sessions had cells with place fields correlated to previous sessions (Figure 6F) and

place field locations did not change as a function of the number of days between

sessions (Figure 6G). In other words, cell reoccurrence decreases across days, but

apparently when a cell does reoccur, its place field remains largely the same.

Conclusion

As previously mentioned, similarities between rat and human physiology, as well

as the increased capacity for complex tasks of learning and memory (Whishaw, 1995;

Frick et al., 2000; Whishaw et al., 2001; Cressant et al., 2007; Rosenfeld and Ferguson,

2014; Hok et al., 2016), position rats as a well-suited model organism for studying

hippocampal function, with several advantages over the mouse. Rats are advantageous

over mice in that, for many diseases and disorders, including those with substantial

hippocampal involvement, their physiology and behavior more closely mirrors that of

humans (Ellenbroek and Youn, 2016). For instance, 5-HT serotonin receptors, which

are prominent in the hippocampus and are implicated in a variety of disorders (including

mood and anxiety disorders, psychiatric disorders such as schizophrenia, addiction and

ADHD), are distributed in rats, but not in mice, in a similar pattern as seen in humans

(Hirst et al., 2003; Ellenbroek and Youn, 2016). Hippocampal neurogenesis, which is

believed to play an important role in learning, memory, and the treatment of depression,

is believed to follow a similar pattern in rats and humans, and is less pronounced in

mice (Snyder et al., 2009). In addition, both pharmacologic and behavioral responses to

addiction are more similar between rats and humans than mice and humans (Jupp et

al., 2013; Parker et al., 2014; Ellenbroek and Youn, 2016).

Having the ability to do calcium imaging in rat CA1 will allow increased research

into these disorders, as well as basic research in learning and memory. Because rats

are substantially larger than mice, they can support larger lens implants and cameras,

which may lead to the ability to record from thousands of hippocampal cells in a single

session during complex tasks. The ability to record a large number of cells across days

will allow research into cell firing changes in the hippocampus, which will provide insight

into more advanced learning and disease mechanisms that are difficult to understand

when only investigating mice. Additionally, the uniquely high percentage of recorded

place cells (>80% of recorded cells) makes CA1 calcium imaging in rats an ideal

method by which to study spatial navigation and learning over days, including

mechanisms behind memory consolidation, spatial remapping, and planning over time.

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Figure captions

Figure 1: GCaMP is expressed in the CA1 cell layer of rat hippocampus

a. Diagram showing injection locations of AAV9-Syn-GCaMP7c, right below the CA1 cell layer. There are four total injection sites, with 0.6uL injected at each site. (CC = corpus callosum, DG = dentate gyrus)

b. Diagram showing level of aspiration in brain tissue. The entire corpus callosum is aspirated until the vertical striations of the alveus are visible.

c. GCaMP expression in the hippocampus, taken with a confocal microscope. The CA1 cell layer is labeled.

d. GCaMP appears to be highly localized to principal cells in the pyramidal cell layer. Left: GCaMP labeling. Center: Labeling for PV+ interneurons using immunohistochemistry (see methods). Right: Overlay of the two images.

e. Diagram of linear track. The animal would run back and forth between the two sides to receive a pellet reward

Figure 2: Hundreds of cells can be captured in a single Ca2+ imaging session

a. Examples of data from two (top and bottom rows) recording sessions. First column: example image from pre-processing. Second column: image post-processing (see materials and methods). Third column: all cell extractions, prior to manual cell sorting. Fourth column: post processing image overlayed with cell outlines identified after manual sorting.

b. Example traces from 15 cells taken during 124 seconds of calcium imaging.

c. Number of cells imaged per session for each rat. Dotted red line indicates the average for each animal.

d. Histogram showing average rate (events/second) for individual cells. Dotted red line indicates the average (0.090+-0.019 events/s)

e. The average calcium event rate (events/second) for individual rats (averages: 0.084+-0.024, 0.074+-0.026, 0.085+-0.020, and 0.116+-0.032 events/second)

f. Correlation of average calcium event rate with average animal running speed.

Average events per second in a session is positively correlated with the

animal's average speed in that session ($p < 0.05$).

g. An example from one session for the cross-correlation at different lags

between the number of calcium events that occur in a second and the

animal's speed. In this example, the maximum cross correlation value at 0s

was about 0.33, and the maximum value occurred at about 0.1s. On average,

the cross-correlation at a lag of 0s was 1.8+-0.08, although the average

maximum cross-correlation was 0.2 +-0.7 at an average lag of +0.47+-0.54s,

indicating that calcium events reliably followed changes in speed, as opposed

to vice versa.

Figure 3: Calcium events are correlated with periods of movement

a. Raster plot of cell firing during linear track running. Top: the animal's speed.

Bottom: raster of peak calcium transients of 352 identified cells.

b. Graph showing an example of speed (plotted in black) overlayed with event rate

(in blue). Bottom panel is a zoom in of a smaller time period.

c. An example logarithmic correlation of animal speed against calcium events/

second.

Figure 4: A high percentage of place cells are recorded on the linear track using Ca2+

imaging

a. Graph showing the percentage of place cells per session for each rat (blue).

Analysis was repeated including only cells with a firing rate >0.01Hz during

movement (red). Dotted lines indicate the average for each animal.

b. Nine example place cells. A firing rate map in both directions of travel is

provided for each cell, with the top map corresponding with running to the

right, and the bottom map corresponding with running to the left. Note the

colorbar scales may be different for each direction of travel. The difference in

mutual information scores between the cell and the top 95% of shuffled data

848	(actual MI-shuffled MI) is indicated in the left corner. If the actual MI was
849	greater than 95% of shuffled data, the number is printed in bold white.
850	c. Distribution of directional mutual information scores after shuffling cells seen
851	in figure 3b. Event data was shuffled 500 times and the top 95% of mutual
852	information scores were determined from shuffled data (red dotted line). If the
853	actual mutual information score (green line) was greater than the upper 95%
854	MI score of shuffled data, the cell was considered to be a place cell.
855	d. Top left: Place field maximum event rate during movement (dotted red line
856	indicates average of 0.15events/s). Top right: average firing rate during
857	movement (red line is average at 0.02events/s). Bottom: Increase from
858	average to maximum rate (max/mean). On average, cells increased firing
859	7.4x from their average firing rate during movement to their maximum firing
860	rate (dotted red line indicates average)
861	
862	**Figure 5:** Place cells are sufficient to decode the animal's position
863	a. Four decoding examples as the animal traverses the track. Black line
864	indicates the animal's actual position, with the stars indicated sample point.
865	The colored circles connected by the red line indicate decoded position. The
866	color of the circles indicates decoding certainty.
867	b. Same as A but decoding is using shuffled units. Decoding is much less
868	accurate using shuffled data.
869	
870	**Figure 6:** Individual cells and their place fields can be maintained across sessions
871	a. Cells can be maintained across sessions weeks apart. Examples from two
872	animals of data taken on the first and final days of recording. Cells are color-
873	matched from the first to last day. Cells in gray represent cells with no
874	matches between the two days. Of note, while the camera was placed in the
875	same approximate position each day, thus allowing landmarks (such as blood
876	vessels) to be followed throughout the study, we did not attempt to perfectly

align landmarks or focal planes across days; instead recordings each day

were made to optimize total cell number in each day as proof of method.

b. Percent of cells that appeared in different numbers of sessions. Symbols

represent averages for individual animals. There was no significant difference

in distribution of values between animals (non-parametric k-sample

Anderson-Darling test, rank statistic: -1.48, $p > 0.05$).

c. Percent of cells that had place fields in different numbers of sessions.

Symbols represent averages for individual animals. There was no significant

difference in distribution of values between animals (non-parametric k-sample

Anderson-Darling test, rank statistic: -1.59, $p > 0.05$).

d. The recurrence probability of a cell based on the number of days between

recordings. Recordings from rat 3 were from found to be from two different

focal planes.

e. Average recurrence probability across days, excluding rat 3. For every day

passing, a cell has a 0.43% less chance of reoccurring.

f. Change in place field location between recording sessions. Included all

sessions recorded for an animal; sessions did not need to be adjacent.

g. A comparison of place field locations for one rat across all six recording

sessions (15 comparisons). For all 15 comparisons using this animal, place

field locations were highly correlated between the sessions. Each point

indicates the place field of a single cell identified in the two session indicated.

Red lines on the graphs indicate a $p < 0.05$ using an F-test. Across all four

animals, we found field locations to be consistently correlated, with 93% of

session comparisons having a significantly positive correlation between place

field locations (f-test, $p < 0.05$).

h. There was no correlation ($p > 0.05$) between r2 value of place field correlations

and days elapsed between sessions.

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Competing Interests

The authors declare no competing interests.

Related Manuscripts

Authors do not have a related or duplicate manuscript under consideration (or accepted)

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Table 1. Comparison of place cell findings in previous studies in mice and rats

Study	Method	Species	Task	Mean spike rate or Ca event rate	% place cells	How place cells were determined
Guzowski et al. 1999	IEG	Rat	Cylinder and platform exploration	NA	**38-39%**	ARC detection
Witharana et al. 2016	IEG	Rat	Environmental exposure and circular track	NA	**22-73%**, mean 33±18%	Homer1a and ARC detection
Talbot et al. 2018	Ephys	Mouse	Open field	1.14±0.12 spikes/s for all pyramidal cells 1.49±0.15 spikes/s for place cells 2.61 ± 0.2 spikes/s mean in-field rate	**47.8%**	overall discharge rate was 0.1-2 AP/s, the spatial coherence was >0.4, and the information content was >0.4 bits/spike
Jun et al. 2020	Ephys	Mouse	Open field and linear track	~0.8 spikes/s mean rate ~3.5 spikes/s maximum rate	**50%**	Mutual information >95% of shuffled data
Maurer et al. 2006	Ephys	Rat	Circular track	Not noted	**~43%** on smaller track **~68%** on larger track	place field has a single cycle (360%) phase precession
Wirtshafter and Wilson 2020	Ephys	Rat	Double sided T maze	~0-5hz in field	**75.2%** exceeded 0.8 bits/spike **63%** had at least one place field and exceeded bits/spike threshold	1) Bits/second >0.8 2) firing in a region must be greater than two standard deviations greater than the mean firing rate and place field area must be at least 15 cm long
Fenton et al. 2008	Ephys	Rat	Forage for food in cylinder and chamber	0.91-1.01 spikes/s depending on environment	**64%** in smaller cylinder **72%** in 6x larger chamber	cell coherence was >0.25 and information content was >0.5 bits/AP, and if the cell had at

				2-2.5 spikes/s mean in-field rate		least one place field.
Markus et al. 1994	Ephys	Rat	8 arm radial maze forced choice	0.56 spikes/s in young animals	**37%**	Bits/spike cutoff of 2.29
Rich et al. 2014	Ephys	Rat	48m long track	Up to ~0.7 spikes/s during running, highly correlated with # of place fields	**65%** at 48m (range 46-77%)	A place field was defined as >=15 contiguous cm of rate map in where firing rate > 2 Hz.
Meshulam et al. 2017	2p calcium imaging	Mouse	VR linear track	Not noted	**~30%,** range 28% to 41%.	Mutual information above 80% of shuffled
Ziv et al. 2013	1p calcium imaging	Mouse	Linear track	~0-0.5 events/s	**20%**	Mutual information above 95% of shuffled
Stefanini et al. 2020	1p calcium imaging	Mouse	Open field	0.06 ± 0.04 ev/s	MI mean+3σ: **~4%** for all cells **~10%** with 0.1ev/s cutoff >95% shuffled MI **~12%** for all cells **~22%** with 0.1ev/s cutoff	Mutual information > mean+3σ of the shuffled distribution. Also compared to mutual information above 95% of shuffled
Kinsky et al. 2018	1p calcium imaging	Mouse	Square and octagon tracks	Not noted	Unspecified	1) had >= 5 calcium transients during the session 2) Mutual information above 95% of shuffled
Kinsky et al. 2020	1p calcium imaging	Mouse	Continuous alternation	0.02-0.03 ev/s (mean ~40-50 events in ~30m session)	Unspecified	1) had >= 5 calcium transients during the session 2) Mutual information above 95% of shuffled

This work	1p calcium imaging	Rat	Linear track	0.02events/s during movement only 0.15events/s mean maximum in-field rate	>95% shuffled MI **77.3%** for all cells **88.3%** with 0.01ev/s cutoff MI mean+3σ: **64.3%** for all cells **73.4%** with 0.01ev/s cutoff	Mutual information above 95% of shuffled Also computed mutual information > mean+3σ of the shuffled distribution.

Figure 1

Figure 2

A

pre-processing | post-processing | extraction outputs | cell overlay

fluorescence (a.u.)

B

Δ F/F

0   Seconds   124

C

Rat 1 | Rat 2 | Rat 3 | Rat 4

Number of cells

Recording day

D

Number of cells

Average events/s

E

Events/S

Rat

F

Average speed (cm/s)

$y = 180x + 21$

Average Events/s Per Session

G

Cross-correlation

Events/s vs Speed

Lag (Sec.)

Figure 3

Figure 4

Figure 5

Figure 6