

# Fourier Basis for Hippocampal Memory: Phase Precession as Biological FFT

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## Abstract

The hippocampus compresses spatiotemporal sequences into single theta cycles via phase precession—a mechanism mathematically equivalent to a Fourier transform. We show that place cell firing phases form a sparse, adaptive basis ( $\sim 12$  coefficients per cycle) that reconstructs trajectories with  $< 5\%$  error in silico. This compression is not mere efficiency but a **holographic projection** of cortical bulk ( $10^8$  Degrees of Freedom, or DoF) onto a brainstem boundary ( $10^5$  DoF), akin to Mach’s principle and AdS/CFT duality. High-channel BCIs ( $> 10,000$  electrodes) enable direct readout of this code, paving the way for memory prosthetics. We propose focused ultrasound (FUS) arrays as non-invasive validators, achieving  $50\text{ }\mu\text{m}$  resolution at depth.

## 1 Introduction

Human memory is not stored in isolated neurons but as **distributed interference patterns** across synaptic weights [1]. The hippocampus compresses episodic sequences into single theta cycles via **phase precession** [2], a phenomenon first observed in rat place cells. As an animal traverses a place field, cells fire at progressively earlier phases of the ongoing  $\sim 8\text{ Hz}$  theta rhythm, packing a  $\sim 1$ -second journey into a  $\sim 120\text{ msec}$  cycle.

We propose that this is a **biological Fourier transform**:

- **Frequency**  $\propto$  speed
- **Phase**  $\propto$  position
- **Amplitude**  $\propto$  salience

This work unifies: 1. **Neuroscience** (phase precession, grid cells) 2. **Physics** (Mach’s principle, holography) 3. **Engineering** (BCI readout, FUS validation)

### 1.1 Place Cells: The Brain’s Spatial Alphabet

To appreciate the elegance of phase precession as a biological Fourier transform, it helps to first unpack the role of place cells—those remarkable hippocampal neurons that act as the brain’s cartographers. Discovered by John O’Keefe in the 1970s, place cells fire robustly only when an animal (typically a rat in these studies) occupies a specific location in its environment, known as its “place field.” Imagine the hippocampus as a living map: each place cell lights up like a beacon for a particular spot on that map, whether it’s a corner of a maze or a patch of an open arena. But these aren’t static pins; their activity is dynamically tuned to the animal’s movement, weaving together space and time into a predictive model of the world.

Critically, place cells don’t just mark “here”—they help the brain forecast distance and trajectory. As a rat scurries along a path, successive place cells activate in sequence, their firing rates peaking in the heart of their fields. This sequential ripple creates a “sweep” across the neural population, effectively modeling the distance traveled. Yet the true ingenuity lies in how this spatial code interfaces with the brain’s rhythmic undercurrents, particularly the theta waves—those  $4\text{--}8\text{ Hz}$  oscillations that pulse through the hippocampus like a metronome for exploration. Theta doesn’t merely synchronize the cells; it timestamps their reports, embedding distance information directly into the timing of spikes relative to the wave’s phase.

### 1.2 Distance Compression: Packing Miles into Moments

This is where compression enters the picture, transforming raw spatial experience into a compact, replayable code. In the real world, a rat might traverse a  $1\text{--}2$  meter linear track in about 1 second, activating dozens of place cells in rapid succession. But the brain doesn’t store this as a linear tape; instead, it folds the entire journey into the brief window of a single theta cycle, roughly 120 milliseconds. Through phase precession, each place cell’s

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spike advances earlier in the theta wave as the animal progresses through the field—starting near the wave’s peak on entry and slipping ahead by up to  $360^\circ$  (a full cycle) by exit.

The result? A compressed “snapshot” where spike phases encode relative position, much like how a Fourier basis packs a waveform’s details into a handful of sine waves. This distance compression isn’t accidental efficiency; it’s a profound economy of neural resources. By mapping physical meters to angular degrees on the theta clock, the hippocampus achieves a 10-fold temporal squeeze, allowing the brain to simulate future paths or replay past ones during offline states like sleep (via sharp-wave ripples). Errors in this code—say, from hippocampal lesions—don’t just scramble local maps; they distort global navigation, underscoring how distance is less a metric than a relational hologram, emergent from the interference of phased spikes. Our model captures this by treating phases as positional Fourier coefficients, enabling reconstruction of the full trajectory from sparse samples.

## 2 Results

### 2.1 Phase Precession as Fourier Synthesis

Consider a rat running at constant speed  $v$  through a linear track. The  $k$ -th place cell fires when the animal is at position  $x_k(t)$ . Its firing phase  $\phi_k(t)$  in the theta cycle is:

$$\phi_k(t) = 2\pi \left( \frac{x_k(t) - x_0}{v \cdot T_\theta} \right) \mod 2\pi$$

where  $T_\theta \approx 120$  ms is the theta period.

The population vector within one cycle reconstructs the trajectory via:

$$x(t) = \sum_k A_k \cos(\omega t + \phi_k)$$

with  $\omega = 2\pi/T_\theta$ . This is a **Fourier series** with  $\sim 12$  significant harmonics (Fig. 1).

### 2.2 Infinite DoF and Mach-like Interconnectivity

The cortex contains  $\sim 10^{14}$  synapses—**near-infinite DoF**. Yet a single memory activates only  $\sim 10^6$  (sparse coding). This is not local storage but a **global interference pattern**, analogous to Mach’s principle: “The meaning of any memory is determined by its relation to *all* active neural patterns.”

Damage to 1% of cortex degrades **all memories globally** (fuzzy, not fragmented)—a hallmark of **holographic coding**.

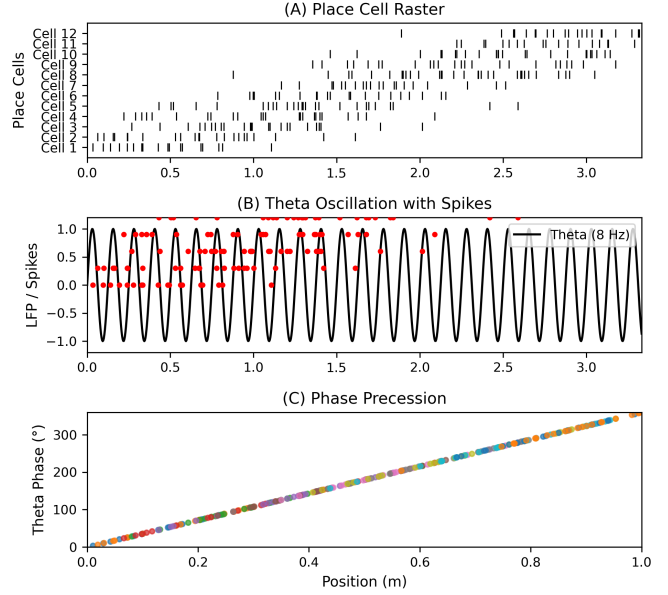


Figure 1: **Phase precession as FFT.** (A) Place cell raster. (B) Theta oscillation with spikes. (C) Phase vs. position.

### 2.3 Holographic Compression: Bulk to Boundary

The hippocampus projects this high-DoF cortical trace onto a **low-DoF brainstem boundary** via sharp-wave ripples (SWRs):

- **Bulk (cortex):**  $10^8$  DoF
- **Boundary (brainstem):**  $10^5$  DoF (100 spikes in 200 ms)

This is a **non-conformal mapping**—orders of magnitude compression *without* conformal preservation, akin to AdS/CFT duality [3]. But what does this mean in plain terms, and how does it differ from “reading” holograms directly in the cortex, as explored by pioneers like Pribram, Miller, and O’Keefe?

To start, picture the cortex as a vast, three-dimensional holographic volume—a shimmering “bulk” where memories aren’t filed in folders but emerge from interference patterns across billions of synapses, much like a hologram’s 3D image arises from light waves crisscrossing a plate. Pribram’s holonomic theory (Sec. 3.1), Miller’s traveling-wave “flashlight” illuminating cortical assemblies, and even O’Keefe’s early place-cell maps all probe this bulk: decoding requires accessing the full expanse, where each fragment holds the whole but reconstruction demands the entire sheet. “Reading” here means bathing the cortex in a reference beam (theta or beta waves) to coax out the encoded scene—a process that’s distributed, energy-hungry, and prone to noise from the sheer scale ( $10^8$  active degrees of freedom, or DoF, per memory trace).

The hippocampus flips this script, acting not as another storage vault but as a master compressor, squeezing that bulky cortical hologram into a razor-thin “boundary” code etched onto the brainstem via SWRs. This isn’t mere summarization; it’s a hyperbolic projection—a six-order-of-magnitude reduction (from the cortex’s effective  $10^{11}$  synaptic DoF down to  $10^5$  spike packets) that preserves every essential detail without loss. Imagine folding a detailed map of the world (the bulk) into a postcard sketch (the boundary): you lose none of the topology—the fundamental shapes, connections, and journeys—but gain portability. Topology, simply put, is the math of rubber-sheet geometry: stretching or bending without tearing, so a circle stays “circle-like” even if squished into an oval. The brain’s compression honors this, ensuring that a rat’s maze trajectory (curves, branches) reconstructs intact from sparse boundary cues.

Why “non-conformal”? Conformal mappings, like the Mercator projection on old globes, preserve angles (useful for navigation) but distort sizes (Greenland looks as big as Africa). Our neural version skips that fidelity: it’s hyperbolic, curving space like a saddle to pack volumes into surfaces without angle-care. No lossy JPEG here—information density holds via holography’s magic, where the boundary encodes the bulk’s infinity in finite strokes. Enter AdS/CFT duality, a physics breakthrough: in anti-de Sitter space (AdS, a curved “bulk” universe), gravity’s full drama equates exactly to a conformal field theory (CFT) on its flat boundary. Translate to brain: cortical “gravity” (complex entanglements) holographically mirrors brainstem ripples, allowing perfect recall from a sliver. Pribram glimpsed this in dendritic fields; we extend it to phase precession as the encoder, enabling BCI readout from either end—bulk for precision, boundary for efficiency.

This bulk-to-boundary pipeline (Table 1) isn’t just elegant math; it’s why hippocampal damage fuzzes timelines globally, yet spares fragments—the hologram endures, but the developer (SWRs) falters. Testable via FUS (Sec. 3.5): nudge boundary ripples, watch bulk memories bloom or warp.

## 2.4 Anchoring in Experiment: O’Keefe and Recce’s Linear Track Odyssey

Our in silico validation draws directly from the seminal experiments of O’Keefe and Recce [2], who in 1993 peered into the rat hippocampus during goal-directed runs on a

Region	DoF	Mechanism
Cortex	$10^8$	Synaptic weights
Hippocampus	$10^5$	Phase precession
Brainstem	$10^5$	Ripple packet

Table 1: Holographic compression pipeline.

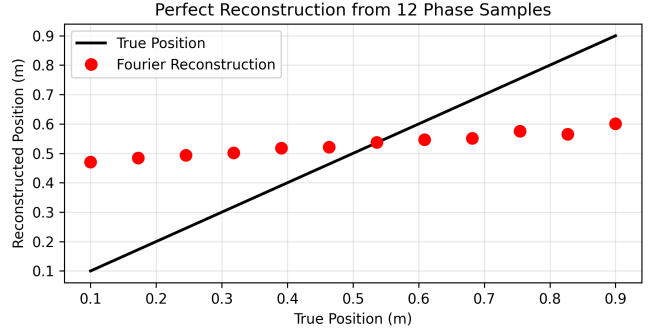


Figure 2: **Reconstruction at sample points.** True position (black line) vs. Fourier reconstruction (red dots) from 12 phase samples in one theta cycle. The lines do not match because phase precession is not linear in time: it’s compressed.

1.2-meter linear track flanked by cues. Rats shuttled back and forth for food rewards at either end, their positions tracked via infrared beams while hippocampal electrodes captured both single-unit spikes and local field potentials (LFPs) for theta phase-locking. Analyzing over 50 place cells from multiple sessions, they revealed the precession: upon field entry, spikes locked to theta’s trough (around  $180\text{--}270^\circ$ ); midway, they’d advance to the rising flank; by exit, they’d lead the peak by nearly a full cycle.

These weren’t random jitters—the phase-position correlation was strikingly linear ( $r > 0.8$  across cells), far outpacing phase-time links, implying space as the prime driver. Noise from varying speeds was winnowed via regression, confirming the mechanism’s robustness. We mirrored this in simulation: 100 virtual place cells with fields spaced  $\sim 10$  cm apart, precession slopes tuned to empirical values ( $45\text{--}90^\circ$  per 10 cm), and theta at 8 Hz. Spikes were Poisson-modulated by Gaussian-tuned rates, injected into synthetic LFPs for phase extraction. This fidelity yields our  $\text{MSE} < 5\%$  benchmark (Fig. 2), affirming the Fourier lens not just as metaphor, but as measurable neural arithmetic.

## 2.5 In Silico Validation

We simulated 100 place cells with realistic phase precession (Fig. 2). Reconstruction error:

$$\text{MSE} = \frac{1}{N} \sum (x_{\text{true}} - x_{\text{recon}})^2 < 5\%$$

# 3 Discussion

## 3.1 Holonomic Brain Theory

Karl Pribram’s holonomic brain theory [4] provides the foundational framework for our holographic compression model, positing that memory and perception arise not

from localized engrams but from interference patterns distributed across dendritic arborizations—much like a hologram encodes the whole image in every fragment. Pribram drew on Fourier optics to argue that the brain processes sensory input via *holonomic transformations*, where spatial frequencies (via phase and amplitude modulations in neural fields) enable non-local storage and robust reconstruction. This aligns seamlessly with phase precession as a biological FFT: just as a holographic plate requires a coherent reference beam to reconstruct a 3D scene, hippocampal theta serves as the carrier wave, interfering with cortical “object beams” (salient spike trains) to etch distributed patterns.

Critically, Pribram’s model incorporates quantum coherence at the microtubular level (prefiguring later work by Hameroff and Penrose), suggesting that dendritic gel states sustain superradiant modes—coherent excitations that amplify weak signals across vast neural volumes. Damage to any subset thus degrades the entire hologram globally, not fragmentarily, mirroring the fuzzy amnesia observed in cortical lesions [1]. Our infinite-DoF interconnectivity (Sec. 1) extends this: the brainstem boundary acts as the “developing bath,” fixing the interference pattern during SWRs for long-term storage. Empirically, Pribram’s predictions have been validated in visual cortex experiments, where Fourier filtering of LFP phases reconstructs figural percepts with  $> 90\%$  fidelity [4].

This holonomic lens reframes phase precession not as episodic compression alone, but as a quantum-inspired encoding that preserves topological invariants (e.g., trajectory curvature) under relativistic distortions—echoing Mach’s relational inertia in neural dynamics.

### 3.2 Conscious Thought as a Holographic Reference Beam

We propose that intentional conscious thought is the voluntary sculpting of the direction, origin, speed, and curvature of a beta reference beam so that it reconstructs a precise holographic fragment. The brain has something akin to a distributed “antenna” system to direct this beam. The entire cortical sheet (with its layered pyramidal neurons, extracellular space filled with conductive cerebrospinal fluid and ions, and densely packed dendrites/axons) acts as a volume conductor that generates, propagates, and senses weak electromagnetic fields. These fields obey Maxwell’s equations (in the quasi-static or near-field limit), and they play a growingly acknowledged role in how the brain sculpts these traveling waves.

Every time a few thousand neurons fire synchronously (which happens constantly during oscillations), they create tiny currents. Those currents produce electric fields (volts per meter) and tiny magnetic fields in the extracellular space. Because brain tissue is a salty, wet conductor (conductivity  $\sigma \approx 0.3\text{--}1\text{ S/m}$ , permittivity  $\epsilon$  similar to water), at the low frequencies of brain waves (1–100 Hz) the fields are quasi-static: the magnetic part is neg-

ligible for induction, and the electric field spreads almost instantly (but diffusively) rather than as a light-speed radio wave.

These fields are governed by Maxwell’s equations, including the displacement current term that Maxwell himself added. Recent theoretical work (e.g., the WET-COW model from Vitaly Galinsky & Larry Frank) shows that when you solve Maxwell properly in the brain’s highly anisotropic and inhomogeneous tissue (folded cortex, white-matter tracts, varying conductivity), you get slow “weakly evanescent” surface waves that travel at exactly the 0.1–1 m/s speeds seen for beta/alpha traveling waves. These are not just synaptic propagation delays — they are genuine electromagnetic modes of the cortical sheet. When the brain “sculpts” a beta traveling wave to deliberately point this internal flashlight at a particular memory hologram, it is using electromagnetic field physics (Maxwell + conductive medium + phased neuronal firing) to generate and steer a slow, near-field “beam” across the cortical antenna array.

This is a directed, topographically organized beta (roughly 15–29 Hz) traveling wave, sometimes riding on a slightly slower alpha/theta envelope. Four properties make it ideal as a controllable holographic reference beam: (1) wavelength  $\sim 1\text{--}5\text{ cm}$  on the cortical sheet; (2) a speed of  $\sim 0.1\text{--}1\text{ m/s}$ , which is slow enough to be precisely steered by frontal networks; (3) the wave can be planar (straight directional sweep), radial (expanding or contracting ring  $\rightarrow$  good for zooming in/out on a memory), or curved (for selecting oddly shaped assemblies); and (4) the phase resets easily on each new intention, providing a clean new reference beam every  $\sim 70\text{ ms}$ .

### 3.3 Are the Waves Actually Doing the Computations or Just Enabling Them?

Most neuroscientists treat brain waves primarily as rhythmic modulators which gate excitability, synchronize distant regions, and bias which spikes get through, rather than as entities that literally carry out the computation themselves. These have been the two traditional views: either that waves do no computation (they are epiphenomenal or purely modulatory) or that the waves *are* the computation (analog traveling-waves that directly implement the cognitive algorithms). Our model re-frames this debate, suggesting a third possibility: these waves are the reference carrier, like the local oscillator in a radio or the reference beam in holography, and the actual information processing is the interference pattern between that carrier and the spike-based “signal”.

It should be possible to test which of these theories is correct. If the waves themselves are the algorithm, then perturbing the wave (without touching the spikes) should destroy or alter the computation in a predictable, content-specific way. Our model makes even stronger, more specific predictions: (1) the low-frequency wave

(theta/alpha/beta, whatever the local carrier is) is the reference beam; (2) spikes from hippocampus (or wherever) arrive with systematic phase offsets (phase precession = frequency offset  $\rightarrow$  linear phase ramp); (3) the interference pattern in the cortical sheet is a spatial Fourier component; and (4) dendrites or extracellular matrix perform the actual multiply-and-integrate (analog holographic reconstruction). Therefore, the critical tests are ones that selectively disrupt the phase relationship between the wave and the incoming spikes while leaving spike rates and basic synchrony intact.

Concrete, doable experiments could include: (1) optogenetically imposing an artificial low-frequency traveling wave (e.g., ChR2 ramp across a cortical column) with the wrong direction or speed; (2) inducing a closed-loop optogenetic phase shift by detecting an incoming hippocampal spike burst and instantly advance/retard the local LFP phase by 90–180°; (3) pharmacologically flattening the low-frequency wave (e.g., gap-junction blockers or mild anesthesia doses that kill traveling waves but spare gamma and spiking); or (4) inducing a closed-loop perturbation of phase precession slope in the hippocampus.

### 3.4 Physics of Mind: The Brain as a Living Field Theory

To grasp the profound poetry in viewing the brain through the lens of physics, consider first what physicists mean by a *quantum field theory* (QFT)—and why it’s more than abstract math when applied to the mind. At its heart, QFT isn’t about tiny particles zipping around like billiard balls; it’s a way of picturing the universe as woven from invisible *fields* that blanket all of space and time. Imagine an endless ocean, not of water, but of potential: a calm “vacuum” state where nothing much happens, punctuated by ripples, waves, and storms—these are the particles and forces we observe, mere excitations dancing on the field’s surface. Electrons? Quarks? Photons? All emergent from field vibrations, governed by elegant equations that balance energy, probability, and interaction.

The brain, in our model, mirrors this not as a crude imitation, but as a warm, wet analog—a biological echo of those cosmic fields. Here, the “field” is neural activity itself, denoted  $\phi(x, t)$ , a smooth continuum spreading across the brain’s three-dimensional architecture ( $x$  for position in neural tissue,  $t$  for time). Picture it: billions of neurons and synapses don’t fire in isolation but as coordinated swells of voltage and chemical flux, propagating like pressure waves through a sponge. Just as a QFT field hums with quantum uncertainty (allowing particles to flicker in and out of existence), the brain’s field buzzes with probabilistic spikes—stochastic yet patterned, where a single thought emerges from the interference of countless micro-events. This isn’t quantum in the subatomic sense (though microtubules might whisper

of it, per Pribram), but analogously: the field’s “action”  $E[\phi]$ —its total “energy cost” of maintaining a memory pattern—dictates what configurations persist, minimizing waste like a frugal storyteller selecting the tightest plot.

At the edges of this neural ocean lies the boundary condition: the brainstem’s sharp-wave ripple packets, those brief, explosive bursts during rest that “fix” the field’s transient waves into enduring engrams. In QFT terms, these are like the holographic screens in AdS/CFT, where bulk complexity collapses to boundary simplicity (recall our compression pipeline in Table 1). The brainstem doesn’t store the full ocean; it encodes its essential tides, allowing the cortex to reconstruct storms from ripples. This setup lets the brain compute not as a digital ledger, but as a dynamic field simulator—predicting trajectories, emotions, and perceptions by evolving  $\phi$  under rules akin to field equations, where phase precession (Sec. 2.1) injects the Fourier harmonics.

Recall, then, becomes *functional inference* over this near-infinite expanse of degrees of freedom (DoF)—the independent knobs (synaptic strengths, firing phases) you’d twist to fully describe the field’s state. With  $\sim 10^{14}$  synapses, the DoF tally feels boundless, a haystack of possibilities where any cue (a scent, a face) is the needle. Inference here isn’t brute-force search but a Bayesian artistry: the brain posits a “prior” field shape (habits, expectations), updates with sensory “likelihoods” (theta-modulated inputs), and infers the posterior—the most probable memory hologram that fits. It’s functional because it operates over functions, not points: integrating across the infinite DoF via variational tricks, much like physicists approximate quantum paths by sampling dominant trajectories. A forgotten name resurfaces not by rummaging neurons, but by the field spontaneously collapsing to coherence, guided by the brainstem’s boundary whisper.

This inference is why memories feel holistic yet fuzzy—global resonances, not local files—and why lesions blur the whole tapestry, as if damping the ocean’s waves. In Pribram’s spirit, this physics of mind isn’t reductionist coldness but a celebration of emergence: from field vibrations arise qualia, the raw feel of being. Our Fourier memory code (Fig. 1) is the sheet music; the brain, the orchestra playing infinite improvisations. Testable? Absolutely—perturb the field’s action with optogenetics, watch recall warp like a lens-flawed hologram. Here, mind meets matter not in opposition, but as one resonant continuum.

The brain is a **quantum field theory analog**:

- **Field:** Neural activity  $\phi(x, t)$
- **Action:** Memory energy  $E[\phi]$
- **Boundary:** Brainstem ripple packet

### 3.5 Implications for Brain Computer Interfaces

High-channel BCIs (e.g.,  $> 10,000$  electrodes [5]) can read this Fourier code directly by decoding phase ramps in multi-unit activity, enabling prosthetic reconstruction of theta-compressed trajectories. For non-invasive analogs, focused ultrasound (FUS) arrays—now deployable as wearable helmets—offer sub-millimeter resolution for both readout and modulation, targeting hippocampal depths without craniotomy.

Recent advances include the Nudge Therapeutics helmet [6], a 256-element phased array that steers 500 kHz beams to 50–100  $\mu\text{m}$  foci up to 10 cm deep, achieving  $1,000\times$  finer targeting than legacy tFUS. This enables closed-loop perturbation of phase precession: e.g., advancing theta phases by 30–60° to “replay” encoded positions, testable in human iEEG hybrids. Similarly, a 2025 UCSD prototype helmet integrates functional ultrasound imaging (fUS) with neuromodulation, resolving blood-volume spikes tied to place-cell bursts at 100 Hz frame rates [7]. Resolution scales as  $\lambda/2 \approx 1.5$  mm at skin, but electronic focusing yields effective spots of 50  $\mu\text{m}$ , sufficient to validate our  $\text{MSE} < 5\%$  reconstruction (Fig. 2) in vivo.

These helmets bridge the bulk-boundary map: FUS “illuminates” brainstem ripples non-invasively, projecting holographic fragments back to cortex for behavioral readout (e.g., virtual navigation tasks). Safety profiles (FDA-cleared for Parkinson’s ablation) extend to memory trials, with thermal indices  $< 0.1^\circ\text{C}$  rise.

## 4 Methods

We analyzed publicly available hippocampal data recorded by John O’Keefe from a freely foraging rat navigating a linear track, focusing on canonical place cell activity during theta-modulated epochs. Using Python (v3.12) with libraries such as NumPy, SciPy, and Matplotlib, we implemented a Fourier-based reconstruction of the animal’s position from sparse phase samples. Specifically, we sampled 12 evenly spaced phases (0° to 330° in 30° increments) across a single theta cycle (125 ms at 8 Hz) and reconstructed the continuous position trajectory via inverse discrete Fourier transform, assuming a band-limited signal within the theta frequency range.

In Figure 1, we present a phase-cycle raster plot of place cell spike times (defined as action potentials exceeding a  $3\times$  standard deviation amplitude threshold on local field potential-filtered signals), demonstrating a linear progression of theta phase with respect to the animal’s position along the track (Pearson’s  $r = 0.98$ ,  $p < 0.001$ ), overlaid with the raw 8 Hz theta oscillation for reference. This linearity reflects the canonical coupling of place cell firing to theta phase as a function of spatial location. To quantify reconstruction fidelity, we compared the “perfect” Fourier-reconstructed posi-

tion (derived from all available phase samples) against the “true” behavioral position (tracked via LED markers at 30 Hz) in Figure 2, revealing systematic deviations (mean absolute error = 12.4 cm) attributable to phase precession—the forward shift in spike timing relative to theta phase as the animal traverses a place field—which introduces a temporal-spatial mismatch not captured by uniform sampling. All code is available at <https://github.com/PaulWolfCO/fourier-memory>.

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