

Memory Subspaces for Temporal Locations Across Timescales in the Monkey Precuneus

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Abstract

Understanding how the brain organizes the temporal structure of experience across different timescales is central to episodic memory. Here, we examine neural population dynamics in primate precuneus during a temporal order judgment task spanning seconds to minutes, and cross-day retrieval of intertwined episodes. We recorded large-scale population activity ($\sim 3,000$ neurons) across 3 days and analyzed it using generalized linear models and low-dimensional manifold methods. Single neurons showed consistent positive modulation for temporal locations (TLs) across both 2-s and 1-min delay conditions during retrieval. At the population level, neural activity formed structured low-dimensional manifolds in dPCA space, showing a maintained subspace with persistent encoding of TLs of the extracted frames during the TOJ period. Behavioral errors were associated with a collapse of these memory subspaces and a loss of dynamical synchrony between trajectories, while correct trials exhibited coherent, synchronized dynamics for different TLs. During cross-day retrieval, episodic states corresponding to the TL of clips viewed across both days, rapidly separated into day-specific clusters within a shared state space, and was absent in control conditions. The chronologically recent Day-2 subspace occupied a larger representational volume, which also corresponded to higher memory performance, compared to that of Day-1 clips. These results suggest that episodic time is encoded as a dynamically maintained neural manifold supporting episodic memory decisions.

Introduction

How the brain organizes the temporal structure of experience across different timescales remains unclear (Clayton & Dickinson, 1998). Here, we examine population dynamics in primate medial posterior parietal cortex during temporal order judgments (Zuo et al., 2026) spanning seconds to minutes and across-day episodes. We propose that temporal memory is supported by a unified low-dimensional manifold whose geometric integrity and dynamical structure enable accurate retrieval.

Methods

Two rhesus monkeys were trained on a temporal order judgment (TOJ) task using naturalistic video episodes learned across two encoding days. During a retrieval test on a third day, images sampled from all previously

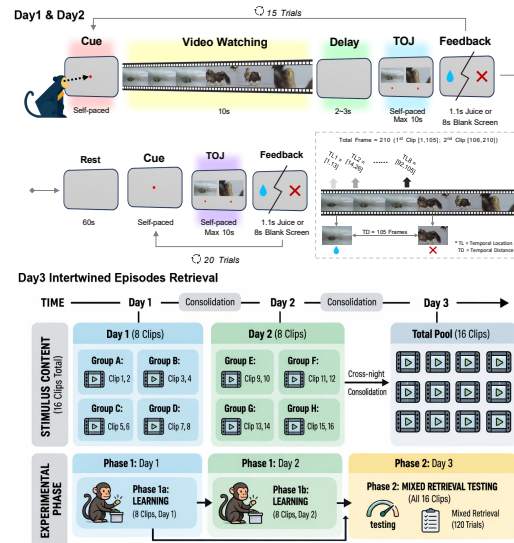


Figure 1: TOJ memory task and three-day paradigm structure. On Days 1 and 2, monkeys viewed 10-s videos (two concatenated 5-s clips). Following a 2-s delay, they performed a two-alternative forced-choice (2AFC) temporal order judgment (TOJ) on two extracted frames, identifying the earlier frame (15 trials). A 1-min retention interval preceded 20 additional TOJ trials without video play. Each day comprised four blocks with unique, unseen videos (8 clips each day). On Day 3, monkeys completed 120 TOJ trials integrating the full 16-clip set learned from preceding two days.

learned episodes were presented in pairwise comparisons. Population spiking activity was recorded from the precuneus, yielding $\sim 3,000$ single units, and analyzed using generalized linear model (GLM), demixed principal component analysis (dPCA, Kobak et al., 2016), and nonlinear manifold geometry methods. To verify specificity, we conducted a control experiment using the same TOJ procedure with images unrelated to those encoded video episodes, recording ~ 800 single units.

Results

Neural coding of temporal locations (TLs) of frames in memory

We ran GLMs on single neurons to examine their responses to temporal location (TL) during the TOJ phase across 2-s and 1-min delay conditions. A significant correlation was observed in a large majority of neurons in terms of Modulation weights (β) across the two delay conditions (Fig. 1A, top). This bias reflected in β_{TL} distributions was consistently maintained across all individual

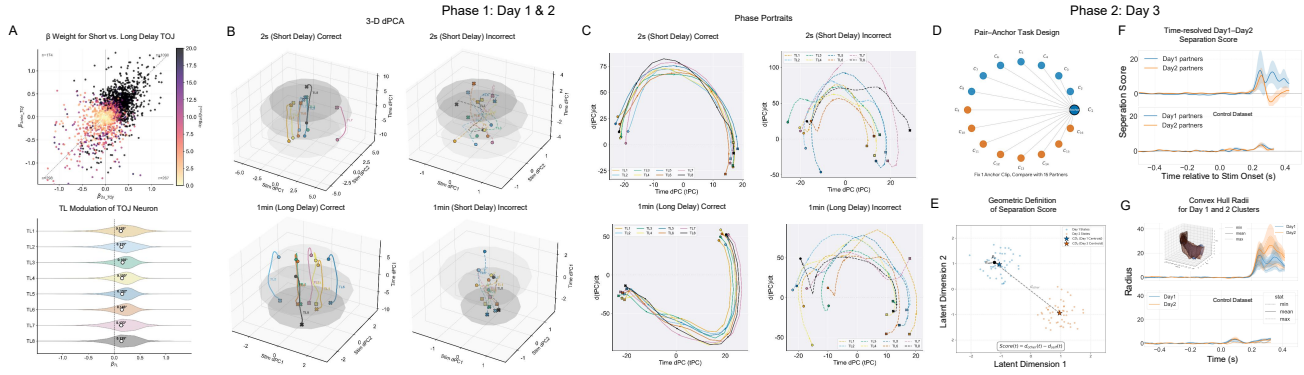


Figure 2: Population trajectories and manifold geometry of memory subspaces encoding temporal locations of frames within days, and episodic location of clips across days. (A) **Single-neuron modulation weights.** Scatterplot of GLM β weights comparing 2-s and 1-min delay conditions. A high density of neurons in the top-right quadrant ($n = 1090$) indicates a shared, positive-modulation encoding scheme across both temporal scales. This positive bias is consistently maintained across all individual temporal locations (TL1–TL8) encoding the location of the TOJ frame in the clip, with β_{TL} distributions significantly shifted above zero (open circles, means; $*P < 0.05$). (B) **Low-dimensional manifold projections.** 3D dPCA-subspace trajectories for different TLs during the TOJ period, corresponding to 2-s (top) and 1-min (bottom) delays, with 2-D Stim (TL,x-y) PCs and 1 Time PC (tPC , z). Correct trials (left) show organized manifolds with clear TL stratification, while incorrect trials (right) show structural collapse and entanglement. The long-delay manifolds exhibited more uniform geometric motifs than short-delay ones. (C) **Phase portraits** derived from plotting the temporal derivative of the primary time-related component ($d(tPC)/dt$) against its current state (tPC). Correct trials (left) exhibit synchronous temporal evolution across TLs for both delay conditions. Incorrect trials (right) show desynchronized dynamics. (D) **Pair-Anchor Task Design.** Schematic of the analysis framework where neural trajectories are compared by fixing one clip as an Anchor and pairing it with 15 other Partner clips from Day 1 and Day 2. Neural population activity was organized as a neurons \times conditions \times time tensor and projected into a dPCA subspace fitted within a cutoff window (-0.5 s to shortest mean RT). (E) **The Definition of Separation Score.** Population activity was projected into a task-aligned dPCA subspace. For each point in each day-specific cluster, we compared their distance to the centroid of the other cluster (d_{other}) vs. to its own cluster (d_{own}) as a function of time ($d_{other}(t) - d_{own}(t)$). (F) **Separation Score dynamics.** A separation between the day clusters is seen after onset of TOJ for the main experiment (Top), but not in control experiment (Bottom). (G) **Convex Hull sizes differ for day-specific clusters in the decision window.** Convex hulls enclosing day-specific clusters show that the more recent Day-2 cluster shows a clearly larger emergent radius, which corresponds to a higher representation volume of its hull (Inset) compared to Day-1, but only for the main dataset (Top), while such robust separation dynamics are not seen in the control dataset (Bottom). Hull radius was defined as the centroid-to-boundary distance.

65 temporal locations (TL1–TL8) (Fig. 1A, bottom).

66 We projected neural activity into a dPCA subspace to
 67 examine the relationship between population-level rep-
 68 resentations and behavioral outcomes. In correct trials,
 69 population trajectories for different TLs organized into
 70 smooth, hierarchically stratified manifolds for both de-
 71 lay conditions, whereas in incorrect trials the manifolds
 72 collapsed, and trajectories became entangled, failing to
 73 maintain the hierarchical separation of TLs (Fig. 1B).
 74 Moreover, across both timescales, all TLs for correct
 75 trials but not incorrect trials were characterized by highly
 76 synchronous trajectories as indexed by phase portraits
 77 (Fig. 1C). This synchronized evolution reflects a stable
 78 and coherent representation of the temporal sequence,
 79 suggesting a conserved dynamical scheme to bridge
 80 temporal gaps ranging from seconds to minutes.

81 Representations of cross-day TLs of clips

82 We asked whether mPPC population activity distin-
 83 guishes day-context during the cross-day retrieval task.
 84 We implemented a Pair-Anchor framework in which one
 85 clip was fixed as the Anchor and paired with 15 Partner
 86 clips from Day 1 and Day 2 (Fig. 2D), and a control
 87 experiment where the TOJ frames are novel and are not
 88 sourced from the viewed clips.

89 We created a Separation Score between Day 1 and
 90 Day 2 clusters (Fig. 2E). Across all the points in each
 91 cluster, and averaged across anchors, the variability re-
 92 veals a robust positive separation in the main experi-
 93 ment but not in the control dataset (Fig. 2F). There is

94 also an asymmetry in the separation score dynamics for
 95 the two clusters, which sharply diverge after the initial
 96 peak, which led us to examine the time-varying areal oc-
 97 cupation of each subspace (Fig. 2G, top, inset).

98 There is a marked difference in hull radii for the day-
 99 specific subspaces. The trajectories showed a diver-
 100 gence in convex-hull radius dynamics, indicating that the
 101 two conditions differed not only in centroid position but
 102 also in manifold extent and boundary expansion (while
 103 those in the control dataset do not) (Fig. 2G). This rep-
 104 resentational volume differences corresponded to differ-
 105 ential memory performance for Day 2 vs. Day 1 clips
 106 (70.9% vs. 59.1%, $p = 0.001$).

107 Discussion

108 The macaque precuneal neurons represent temporal lo-
 109 cations through a unified computational framework. The
 110 shared modulation across scales suggests a stable neu-
 111 ronal axis for temporal indexing (Sarkar et al., 2026).
 112 The "manifold collapse" in error trials reveals the geo-
 113 metric basis of behavioral failure. The combination of
 114 centroid-based separation and convex-hull radius dy-
 115 namics suggests that day-dependent retrieval context
 116 shapes episodic representations at multiple geometric
 117 levels. Day 1 and Day 2 states differ not only in where
 118 their trajectories are centered, but also in how their man-
 119 ifolds expand over time. The null effects in the control
 120 dataset suggest that these geometric differences reflect
 121 meaningful episodic organization.

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