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Age-related brain atrophy and the positive effects of behavioral enrichment in middle-aged beagles

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1 Title Page

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27 Aging dogs serve as a valuable preclinical model for Alzheimer's disease (AD) due to their natural 28 age-related development of beta-amyloid (A β) plaques, human-like metabolism, and large brains that 29 are ideal for studying structural brain aging trajectories from serial neuroimaging. Here we examined the 30 effects of chronic treatment with the calcineurin inhibitor (CNI) tacrolimus or the nuclear factor of 31 activated T cells (NFAT)-inhibiting compound Q134R on age-related canine brain atrophy from a 32 longitudinal study in middle-aged beagles (36 females, 7 males) undergoing behavioral enrichment. 33 Annual MRI was analyzed using modern, automated techniques for region-of-interest -based and voxel-34 based volumetric assessments. We found that the frontal lobe showed accelerated atrophy with age, 35 while the caudate nucleus remained relatively stable. Remarkably, the hippocampus increased in 36 volume in all dogs. None of these changes were influenced by tacrolimus or Q134R treatment. Our 37 results suggest that behavioral enrichment can prevent atrophy and increase the volume of the

ABSTRACT

38 hippocampus but does not prevent aging-associated prefrontal cortex atrophy.

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40 SIGNIFICANCE STATEMENT: Aging canines naturally show significant neuropathological 41 similarities to human aging and AD, making them valuable translational models for testing disease-42 modifying treatments. We applied modern, state-of-the-art longitudinal volumetric analysis approaches to evaluate treatment effects from structural MRI in a large cohort of middle-aged beagles treated with 43 44 the FDA approved calcineurin inhibitor, tacrolimus, or the experimental NFAT inhibitor, Q134R, while 45 undergoing extensive behavioral enrichment. We show increased hippocampal volumes across all dogs, 46 even control placebo dogs, compelling evidence for a strong enrichment-related benefit on hippocampal 47 structural integrity. Our findings are the first of its kind to demonstrate benefits of behavioral 48 intervention on longitudinal structural brain changes in a higher mammalian model of aging and AD.

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50 Keywords: beagle, calcineurin, Q134R, prevention, tacrolimus, Alzheimer's disease, amyloid,

51 neurodegeneration, prevention

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55 Introduction

56 Canines share aging-related neuropathological features with humans, making for a valuable 57 translational model for Alzheimer's disease (AD). Aging canines intrinsically develop diffuse beta-58 amyloid (A β) pathology that is associated with cognitive decline akin to human mild cognitive 59 impairment (MCI) (Cotman & Head, 2008). Furthermore, cognitive functions tested in beagles that are 60 relevant in AD, including spatial memory and executive function, are vulnerable to aging and A β burden 61 (Chan et al., 2002; Head et al., 1998; Rofina et al., 2006; Studzinski et al., 2006; Tapp et al., 2003). Their 62 drug tolerance and metabolism align with humans as well (Martinez et al., 2021). Furthermore, the 63 larger, gyrencephalic brain of beagles compared to lower mammalian AD models proves advantageous 64 for in vivo structural assessments with magnetic resonance imaging (MRI) and for larger sampling 65 volumes for postmortem histological evaluations. These features, coupled with their shorter lifespans 66 relative to humans, make them ideal for studying longitudinal effects of behavioral and pharmacological 67 interventions (Su et al., 2005; Head et al., 2008; Araujo et al., 2022).

68 Clinical trials in AD patients often target the classic pathological hallmarks, the aggregation of A β 69 plaques or tau neurofibrillary tangles (Congdon & Sigurdsson, 2018; Perl, 2010), but have failed to result 70 in effective therapies. Investigations of preventative therapeutics targeting alternative dysfunctional 71 mechanisms are thus warranted (Crous-Bou et al., 2017). A promising target is the hyperactive signaling 72 of the Ca²⁺/calmodulin-dependent protein phosphatase calcineurin and its substrate transcription 73 factor, the nuclear factor of activated T-cells (NFAT). A β plaque aggregation is linked to aberrant 74 calcineurin/NFAT hyperactivity within neurons and astrocytes, leading to neuroinflammation, Ca²⁺ 75 dysregulation, synaptic dysfunction, and excitotoxicity (Norris et al., 2005; Kuchibhotla et al., 2008; 76 Reese & Taglialatela, 2010). This hyperactivity is associated with cognitive dysfunction in transgenic AD 77 models (Sompol et al., 2017) and in early AD patient tissue (Abdul et al., 2009, 2011). Calcineurin 78 inhibition ameliorates these effects in transgenic mouse models of AD (Taglialatela et al., 2009; Stallings 79 et al., 2023), and was previously linked to a profound reduction in dementia incidence among solid 80 organ transplant patients maintained on calcineurin inhibitors (Taglialatela et al., 2015). Our group 81 recently showed that a low dose of the FDA-approved calcineurin inhibitor (CNI) tacrolimus protected 82 against age-related microstructural gray matter changes within the hippocampus, parahippocampal 83 gyrus, and prefrontal cortex of middle-aged beagles compared to placebo after one year (Radhakrishnan 84 et al., 2021). This offers promising evidence for the potential of tacrolimus repurposed as a preventative 85 treatment of AD.

86 Postmortem evaluations in beagles describe typical age-related characteristics across the canine 87 lifespan in great detail (Vite & Head, 2014), but in vivo characterizations are limited. Of the available 88 cross-sectional MRI studies in beagles, findings largely align with postmortem evidence of ventricular 89 widening (González-Soriano et al., 2001; Kimotsuki et al., 2005), early vulnerability of the frontal lobe, 90 (Tapp et al., 2004) and later vulnerability of the hippocampus to neuron loss (Tapp et al., 2004). 91 However, individual differences among aging canines, even within the same breed, parallel the 92 structural heterogeneity of the aging human brain, adding ambiguity to cross-sectional examinations 93 (Cotman & Head, 2008). Furthermore, age and cohort effects pose additional challenges to 94 interpretations of findings (O'Brien, 2017).

- 95 Longitudinal studies are fundamentally immune to such effects and allow for direct assessments
- 96 of individual trends over time. While prior MRI studies in aging canines qualitatively assessed
- 97 neuroanatomy (Gross et al., 2010) or employed visually guided methods for large-scale brain aging
- 98 characteristics (Su et al., 2005), visual inspection alone falls short in detecting subtle alterations during
- 99 aging or assessing long-term interventional effects often measured at the submillimeter level in
- 100 histological evaluations.

Innovative software for analyzing human neuroimaging data have been successfully applied to
 just a handful of canine studies but have regularly required some user intervention for standard
 preprocessing procedures including brain extraction (Milne et al., 2016), or for delineating brain regions
 (Tapp et al., 2004), which can be time-consuming and prone to user error when manually performed.
 Hence, an analytical framework for applying automated and standardized methods for reliable
 longitudinal analysis for canine imaging data is necessary for enhancing reproducibility and translating
 neuroimaging biomarker endpoints to clinical trial outcomes.

To this end, we leveraged advanced, open-source image analysis tools to evaluate structural alterations in a prevention study in healthy middle-aged beagles undergoing chronic CN/NFAT inhibitor treatment and behavioral enrichment. Regional volume changes were assessed using atlas-based volumetry as well as deformation-based morphometry (DBM) from high-resolution T1-weighted imaging collected annually for three years. We hypothesized that aging would be associated with brain region specific losses in volume that may be protected by a combination treatment approach with behavioral enrichment and CN or NFAT inhibitor treatment.

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116 Materials and Methods

117 Experimental Design

118 Animals

119 The study began with forty-five purpose-bred intact adult beagles assessed for general health 120 status and cognition, as previously described (Christie et al., 2008; Head et al., 1998; Milgram et al., 121 1999; Studzinski et al., 2006; Tapp et al., 2003). The final sample included 43 dogs (36 females, 7 males) 122 ranging from 5 to 8.7 years old. The dogs were then divided into three groups. One group was treated 123 with a placebo (n=14) while two groups were treated with a chronic low dose of the FDA-approved CNI 124 tacrolimus (n=15) or the NFAT-inhibiting small chemical compound Q134R (Hackler et al., 2019; Sompol 125 et al., 2021) (n=14) as part of a longitudinal preclinical study investigating their potential for preventing AD-related pathology when treated in middle age (Fig. 1). Tacrolimus was given at 0.075 mg/kg (2X daily, 126 127 P.O.). Dosage for Q134R was increased after the first year of treatment from 4 mg (2X daily, P.O.) to 8 128 mg (2X daily, P.O.) for years two and three. Two dogs required euthanasia prior to reaching the third 129 year of MRI scanning (T3) after their spontaneous development of health conditions unrelated to 130 treatment that were unable to be sufficiently managed with medical intervention (Table 1).

131Blood samples were taken every six months to monitor overall health and potential adverse132effects via assessments of blood urea nitrogen, creatine, and phosphorous levels since tacrolimus can

- 133 cause nephrotoxicity in solid organ transplant patients at higher doses (Randhawa et al., 1997). All
- 134 institutional and national guidelines for the care and use of laboratory animals were followed.
- 135

136 Behavioral enrichment

137 The behavioral enrichment paradigm consisted of daily exercise, socialization, cognitive testing, 138 rotating toys, and social compatible group indoor/outdoor free play. Free play was provided for thirty minutes and in male-only or female-only groups. Sex-matched dogs were pair-housed in the mornings 139 140 and split into their individual kennels in the afternoon prior to feeding. Six sets of two play objects were 141 rotated through each kennel at weekly intervals. Furthermore, dogs were trained and tested on a 142 battery of neurocognitive tests designed to capture age-related decline across several cognitive 143 domains. Briefly, the tasks included spatial learning and memory tasks, landmark discrimination, oddity 144 discrimination, size discrimination, black/white discrimination, as well as reversal learning. Detailed 145 descriptions of these tasks are described by Davis et al (Davis et al., 2017). The dogs were tested at 146 baseline prior to receiving treatment and continuously trained and tested on these tasks 5 days/week 147 (20-40 min depending on the task) throughout the entirety of the study.

148

149 Neuroimaging

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151 IMAGING DATA ACQUISITION

152 Imaging data were collected at the Magnetic Resonance Imaging and Spectroscopy Center 153 (MRISC) at the University of Kentucky in Lexington, KY, USA. Animals were fasted overnight and placed 154 under general anesthesia using propofol (4-8 mg/kg, i.v., by slow injection to effect). Following induction and orotracheal intubation, anesthesia was maintained with 1–4% isoflurane delivered in 100% O2 155 during magnetic resonance imaging (MRI) scanning. As part of the standard protocol, dogs were also 156 157 premedicated with 0.4 mg/kg of Meloxicam as a prophylaxis for CSF collection-induced soft-tissue injury 158 and headache/nausea. All CSF collections were completed post-MRI. A 3T MRI scanner (Siemens Prisma 159 Fit) with a 15-channel transmit-receive knee coil was used to scan the canine brain across 4 time points: 160 at baseline before treatment with placebo, tacrolimus or Q134R (T0), and annually for three years (T1-161 T3). For structural imaging, a high-resolution T1-weighted image was collected using a magnetization-162 prepared rapid gradient-echo (MPRAGE) sequence (repetition time (TR)=2530 ms; echo time (TE) = 2.49 163 ms; flip angle = 7°; matrix size = $0.4 \times 0.4 \times 0.7$ mm; averages = 1; average acquisition time = 10 min, 30 s; 164 voxel size 0.352x0.352mm, slice thickness 0.7mm).

165

166 IMAGE PROCESSING

167 The T1-weighted images were first visually inspected for quality control. No obvious adverse 168 effects of treatment on brain structure were identified for any dogs. Minor ghosting artifacts that did 169 not overlap with the brain were edited out from the 3D images of two dogs to overcome faulty 170 registration during image preprocessing. Next, all images were processed using the Advanced 171 Normalization Tools (ANTs) software (version 2.3.1). ANTs has an extensive track-record of robust

- 172 performance in lifespan analyses of brain morphology (Tustison et al., 2014) and has been recently
- implemented in canine neuroimaging studies for the development of *in vivo* and *ex vivo* group templates
- and digital atlases (Czeibert et al., 2019; Datta et al., 2012; Johnson et al., 2020). The high-resolution
- group template developed for an earlier report on the tacrolimus-treated and placebo-treated dogs was
 used in the present study (Radhakrishnan et al., 2021). For atlas-based ROI analyses, we used a recently
- used in the present study (Radhakrishnan et al., 2021). For atlas-based ROI analyses, we used a recer
 developed high-resolution stereotaxic canine brain atlas with detailed parcellations (Johnson et al.,
- 178 2020) and created a simplified version by grouping cortical parcellations by their respective gyri. This
- 179 simplified atlas, hereby referred to as the "Johnson atlas," was registered to the group template space
- using the affine+SyN algorithm for diffeomorphic image registration. Finally, CSF posteriors were
- obtained from processing the group template through the ANTs cortical thickness pipeline and were
- added to the CSF segmentation map by Johnson et al. (Johnson et al., 2020) which was then binarized
- and smoothed to create a more detailed CSF prior probability map that can better resolve CSF
- 184 compartments within narrow sulci.

185 Individual image processing was then performed using a longitudinal image analysis pipeline 186 newly available within ANTs that has demonstrated superior performance in humans for detecting 187 disease-related structural alterations from serial neuroimaging (Tustison et al., 2019) and interventional 188 effects in clinical trials (Song et al., 2022). Here, minor customizations were made to the pipeline to 189 account for canine brain size and voxel dimensions. An important feature of this pipeline is the creation 190 of an unbiased single-subject template (SST) used as an intermediate reference space between the 191 group template and time point images for reducing registration errors and optimizing sensitivity to 192 longitudinal volume changes (Fig. 1). In the pipeline, each dog's SST was first generated from its set of 193 T1-weighted images that underwent preprocessing including N4 bias-field correction (Tustison et al., 194 2010), brain extraction, and probabilistic tissue segmentation via Atropos (Avants et al., 2011) with 195 reference to the group template, resulting in prior probability maps in the space of the SST as well as 196 non-linear SST to group template warps. Each time point image was then rigidly aligned to the SST and 197 denoised prior to undergoing the same preprocessing steps as the SST while treating the SST as the 198 reference template, resulting in tissue segmentation maps and warps to the SST for each time point 199 image.

200

201 ROI-BASED VOLUMETRY

202 To obtain regional volume measures at each time point, the Johnson atlas was registered to the 203 processed time point images by applying the warps generated from combining the template to SST and 204 SST to time point warps. The volume of each region was calculated in each dog's native time point space 205 as the sum of all voxels within each atlas parcellation with CSF voxels subtracted converted to cubic 206 millimeters. The left and right ROI volumes were adjusted by the log-Jacobian determinant of the linear 207 SST to group template transformation of each animal, a scalar value associated with approximate 208 intracranial volume (ICV) (Buckner et al., 2004) and z-transformed with respect to each ROI's mean 209 baseline volume. We did not observe any significant differences between bilateral ROIs at baseline so 210 mean bilateral volumes were calculated for subsequent analysis.

- 211 Several cortical ROIs and the cerebellum were excluded from further analysis due to erroneous
- tissue segmentations at brain versus non-brain boundaries for several dogs at various time points.
- 213 Brainstem and claustrum ROIs defined within the original Johnson atlas were not included in the
- analysis. A final total of 24 bilateral ROIs were examined that included cortical structures from frontal,
- sensory-motor, cingulate, occipital, parietal lobar areas as well as three subcortical structures:
- amygdala, caudate nucleus, and hippocampus (Table 2).
- 217

218 DEFORMATION-BASED MORPHOMETRY

As a complimentary volumetric analysis, we performed deformation-based morphometry (DBM)

- for a more granular, voxel-based assessment of volume changes. Briefly, maps of the non-linear warps
- 221 generated from voxel-wise registration of each image to a reference image in the same coordinate
- space (e.g., the time point image moving to the SST) can be converted to a log-Jacobian determinant
- 223 map which represents the amount of expanding or contracting of each voxel during the registration.
- 224 When registering to a common template space, the log-Jacobians can be considered a voxel-based
- summary measure of volume that can be compared across individuals (Ashburner et al., 1998). Here, we
- combined the warps generated from the time point to the SST registration and SST to the group
- template registration during preprocessing. Creating "DBM maps" by this method, these maps are set in
- the group template space that capture each dog's longitudinal changes with respect to its SST as well as
- cross-sectional differences with respect to the group template. Each DBM map was resampled to a
- 230 0.5mm isotropic resolution and was slightly eroded by two voxels to exclude problematic edge voxels
- that were found to include CSF, pia, or dura in several dogs' native space (Fig. 1).
- 232

233 Statistical analysis

234

235 ROI VOLUMETRY

First, for an initial evaluation of general atrophy rates regardless of treatment group, we calculated simple ROI volume slopes over time for each dog from bivariate ordinary least squares regressions. We then used a non-parametric rank analysis for evaluating differences between regional atrophy rates using the Friedman test. We performed post-hoc pairwise comparisons of rank-sum differences as well as one-sample t tests against zero on each set of ROI slopes and adjusted for multiple comparisons using Dunn's post-hoc test from the SciPy library in Python.

242 Next, we conducted separate multivariate linear mixed effects (LME) analyses per ROI. LME models can flexibly account for incomplete data, dependency within repeated within-subjects 243 244 measurements, and hierarchically structured data, which can improve the reproducibility of longitudinal 245 studies (Yu, Guindani, et al., 2022). To predict ROI volume, we assigned treatment group, baseline age, 246 years elapsed since the initial baseline scan ("time"), and a time by treatment interaction term as fixed 247 effects, and a random intercept was assigned at the level of individual dogs (i.e. random effect of 248 subject). The p-values for the LME parameter estimates from each ROI analysis were adjusted for 249 multiple comparisons using the Holm-Sidak method for a family-wise error rate of 5%.

250 Lastly, we used an all-ROI-inclusive LME model by adding "region" as a predictor variable to evaluate differential patterns of volume changes in a single model using the hippocampus as the 251 252 reference level. The region with the least decline observed from the two prior analyses was used as the 253 reference ROI in the model. Since the per-region LME models did not show reliable evidence for 254 treatment by time interactions (Fig. 4), the treatment predictor was excluded from this LME analysis to 255 evaluate sample-wide trends in volume changes regardless of treatment group. Thus, the fixed effects 256 for this model included region, time, z-transformed baseline age, a time by region interaction term, and 257 a random intercept was again assigned per dog. Here, the estimates for baseline age and time attributed 258 to the reference region, and the estimates for each time by region interaction are thus conditional on 259 the estimated effect of time. Diagnostic model fit evaluations using the Akaike Information Criterion 260 (AIC) and the Bayesian Information Criterion (BIC) indicated that a model excluding treatment was

- 261 indeed preferred over a model that included interactions with treatment.
- 262

263 DBM ANALYSIS

264 We conducted a brain-wide linear mixed effects analysis of voxel-wise volume change on the DBM maps using the 3dLMEr package in the Analysis of Functional NeuroImages (AFNI) software suite 265 (Chen et al., 2013). Here, treatment group, baseline age, time, and a time by treatment interaction term 266 267 were assigned as fixed effects, and a random effect was assigned per dog for predicting the log-Jacobian 268 determinant (i.e., volume) per voxel. Baseline age values were centered on the mean baseline age of 6.7 269 years (age – mean age) and time was centered on baseline year (time=0). Minimum cluster size 270 thresholds for the Z-statistical maps of each main effect, interaction, and for treatment contrasts over 271 time (tacrolimus over Time vs. Placebo over Time, Q134R over Time vs. Placebo over Time), correcting 272 for multiple comparisons, were empirically determined from 100 Monte Carlo simulations in 3dLMEr. 273 Baseline age and treatment were permuted across subjects and time was permuted within subjects. 274 Each critical cluster size was determined from the 5% probability of the largest resulting false-positive 275 clusters generated using AFNI's 3dClusterize with a voxel-wise threshold of p=0.0001.

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277 Code availability

278 Code for data processing and analysis is available at https://github.com/StarkLabUCI/canine-long-sMRI.

- 279
- 280 Results

281 Regional selectivity of volume losses, hippocampal volume increases

First, rank analyses of the dog-specific slopes of volume over time revealed consistent patterns of cortical volume loss among 17 cortical ROIs (Fig. 2a,b, Table 2). All frontal lobe ROIs, including the pregenual, proreus, and orbital gyri, were ranked highest for exhibiting the most rapid rates of volume loss (orbital gyrus: median slope (z)=-0.17, median rank=5 out of 24; gyrus proreus: median slope=-0.21, median rank=5; pregenual gyrus: median slope=-0.15, median rank=6) (Fig. 3) and declined significantly faster than ten other regions (Fig. 2b). Furthermore, the posterior compositus and sylvian gyri within the temporal lobe ranked next highest for rapid volume loss (posterior compositus gyrus: median slope=-

289 0.18, median rank=6; sylvian gyrus: median slope=-0.14, median rank=8). There was no apparent

290 evidence of volume loss from the simple volume slopes of the genualis, suprasylvian, parietal coronal,

and entolateral gyri, or from the amygdala (Table 4). Most notably and contrary to typical brain aging in

canines (Tapp et al., 2004; Su et al., 2005; Kimotsuki et al., 2005), the hippocampus showed strong

evidence of unique volumetric increases over time (median slope=0.16, median rank=24) and the
 caudate nucleus also showed evidence for a mild volumetric increase (median slope=0.06, median

rank=22) (Fig. 2a,b, Fig. 3, Table 4). The remaining regions in the cingulate, sensory-motor, occipital, and

296 parietal lobes also showed volume loss but to a lesser extent (Fig 2a, Table 4).

297

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299 Next, we employed separate multivariate LME models per ROI to test for longitudinal effects of 300 treatment and baseline age on volume losses as a function of time. We did not observe any significant 301 time-by-treatment interactions for tacrolimus or Q134R for any ROIs after adjusting for multiple 302 comparisons. The regions that showed a significant main effect of time on volume loss were largely 303 consistent with the regions found to decline in the above simple slope analyses. However, we found that 304 the LME models for the ectosylvian, parietal marginal, recurrens, and proreus gyri, and the caudate 305 nucleus did not show sufficient evidence of any volume changes over time (Fig. 4). A strong increase in 306 hippocampal volume over time across all groups was also evident by LME analysis (β =0.154, SE=0.034, 307 adjusted p<0.001, CI[0.088, 0.220]).

308 The noteworthy increases in hippocampal volume across all groups motivated the use of the 309 larger, all-ROI-inclusive LME analysis to directly contrast its rate of decline to that of each ROI in a single 310 model. Here, the parameter estimates for each ROI are conditional upon the main effects of time and 311 age, with the volume of the hippocampus as the reference region (Fig. 5a). Treatment group was 312 excluded from this model due to the lack of evidence for treatment effects in every per-region LME 313 analysis. We found volume increases in hippocampal volume over time as we observed in our prior 314 analyses (β =0.181, SE=0.048, p<0.001, CI[0.087, 0.275]), which equates to a 1.74% average annual 315 increase in volume. Critically, every time-by-region interaction apart from the caudate nucleus showed 316 distinct rates of volume decline relative to the hippocampus, underscoring the unique trajectory of 317 hippocampal volume changes over the three years. Furthermore, the differential pattern of volume 318 decline across the other ROIs were highly consistent with the simple slope analysis, specifically the 319 accelerated volume decline of frontal pregenual, proreus, and orbital gyri and temporal posterior 320 compositus gyrus (Fig. 5a,b). The full set of parameter estimates for this model can be found in Table 5.

321

322 Characterizing whole-brain spatiotemporal patterns of volume changes using DBM

Atlas-based approaches to assessing volumetry have their clear merits but run the risk of missing patterns or effects that do not correlate well with region boundaries, which motivated the use of whole-brain DBM. Thus, we conducted a voxel-wise linear mixed effects analysis of the log-Jacobians or "DBM maps" that were generated from spatial registrations of the time point images to the SST and group templates. We used a strict voxel-wise threshold of p=0.0001 to identify clusters of the most
 extreme volume changes predicted by time, baseline age, and treatment by time interactions.

Consistent with the ROI-based LME analyses, post-hoc linear contrasts examining differences from placebo for time by treatment interactions (tacrolimus-placebo minimum cluster size threshold = 29 mm³; Q134R-placebo minimum cluster size threshold = 15 mm³) did not yield any significant clusters

of treatment-related effects on volume change by tacrolimus or by Q134R (data not shown).

333 Next, we wanted to examine cross-sectional aging effects on regional volume in the brain 334 independent of longitudinal treatment effects. A cluster analysis of the log-Jacobians revealed five 335 clusters for the main effect of age on volume decline (log-Jacobian<0) on all treatment groups with time 336 fixed at baseline (minimum cluster-size threshold = 25 mm³). We found evidence of focal volume loss 337 with every unit increase in age that was mainly posteriorly localized in the brain (Fig. 6a-d). The largest 338 cluster of reduced volumes was found in the occipital marginal gyrus, medial occipital gyrus, and 339 posterior cingulate gyrus in the left hemisphere (53 mm³) (Fig. 6c), and the next largest cluster 340 encompassed only the occipital marginal gyrus and the medial occipital gyrus in the right hemisphere 341 (47 mm³; Fig. 6c). A smaller cluster spanning the occipital and temporal lobes was found within the left 342 suprasylvian and ectosylvian gyri (35 mm³; Fig. 6b) and similar-sized cluster was found within the 343 interthalamic adhesion (19 mm³; Fig. 6d). Lastly, a cluster within the left occipital lobe spanning the left 344 inferior medial occipital gyrus and pararecurrens (16 mm³) had the most extreme age-related volume 345 loss at baseline (Fig. 6a).

346 Next, to assess the volumetric changes over three years in the study which included behavioral enrichment across all groups, we performed a cluster analysis for the main effect of time from the DBM 347 348 maps (minimum cluster-size threshold=10 mm³) (Fig. 7). Our DBM findings corroborated our ROI-based 349 results where cortical shrinkage (log-Jacobian<0) was the most prominent in frontal and temporal lobes. 350 Furthermore, the greater spatial precision of this voxel-based volumetric estimation identified hotspots 351 of volume loss within the posterior cingulate gyrus and the suprasylvian gyrus. The segmentation-352 agnostic DBM maps also allowed us to assess areas not defined by the Johnson atlas or ROIs that were 353 previously excluded due to segmentation issues. We found rapid annual volume loss within the gyrus 354 subproreus of the frontal lobe, the precruciate, pre- and postcentral gyri of the sensory-motor lobe, as 355 well as significant volume loss within the genu, body, and splenium of the corpus callosum, and within 356 midbrain and brainstem areas. Cerebellum clusters not shown.

357

358 Conversely, areas of significant expansion (log-Jacobian>0) were observed over the lateral 359 ventricles indicating ventricular widening, consistent with typical canine aging (Gunde et al., 2020; Su et 360 al., 2005) (Fig. 7). The hippocampus had significant expansion along most of the dorsal hippocampus, 361 but the ventral hippocampus was largely unchanged (Fig. 8). Furthermore, we found that the third 362 ventricle immediately adjacent to dorsal hippocampus showed extreme volume loss which is consistent 363 with dorsal hippocampal volume expansion. Lastly, we observed unexpected expansion across occipital 364 and frontal white matter areas consistent with experience-dependent white matter plasticity (Mendez 365 Colmenares et al., 2021) (Fig. 7).

367 Discussion

In this study, we investigated cortical and subcortical volume changes in aging beagles, a natural model of AD (Cotman & Head, 2008; Vite & Head, 2014), undergoing long-term behavioral enrichment and tacrolimus, Q134R, or placebo treatment beginning in middle age. Our main goal was to explore the spatiotemporal patterns of brain atrophy in normal canine aging and assess the potential neuroprotective effects of CN/NFAT inhibitor treatment on brain volume. Using segmentation-based ROI

- volumetry and whole-brain, voxel-based DBM, we uncovered striking and unexpected evidence of
- increased bilateral hippocampal volume in all three groups across three years.

375 Our ROI-based evaluations showed that total hippocampal volume increased at an average rate 376 of about 1.74% per year across treatment groups, contrasting with the age-related hippocampal volume 377 decline observed in previous cross-sectional neuroimaging studies in laboratory beagles (Tapp et al., 378 2004). We did not observe any relationships between age and hippocampal volume at baseline prior to 379 the behavioral enrichment protocol (Fig. 9), suggesting that the hippocampal volume increases were 380 likely not a feature of typical hippocampal maturation in beagles. Instead, we argue that these increases 381 may be attributed to the high levels of behavioral enrichment in the present study that included social 382 interaction, exploration, physical exercise, and sensory stimulation, all of which are known to induce a 383 number of neurobiological changes. Previous studies showed behaviorally-enhanced adult neurogenesis 384 in canines within the hilus of the dentate gyrus (DG) following behavioral enrichment (Siwak-Tapp et al., 385 2008) and a dorsal-ventral gradient in neurogenic potential in the canine hippocampus (Bekiari et al., 386 2020). Although in the present study our DBM results showed volumetric increases that were mainly 387 localized to dorsal hippocampus, whether these changes were primarily attributed to hippocampal 388 neurogenesis is limited to speculation given the lack of a non-enriched control group in this study. 389 Furthermore, other exercise-related angiogenic mechanisms such as vascular plasticity and increased 390 cerebral blood volume may drive up hippocampal volume as well (Kim et al., 2021), which has been 391 observed with high-resolution MRI in human adult exercise intervention studies (Maass et al., 2015). 392 Whether angiogenesis and neurogenesis are interdependent mechanisms at play in the hippocampus, or 393 in other regions that possibly undergo adult neurogenesis such as the hypothalamus (Fowler et al., 394 2008), remains to be explored. A key aim of our histological analyses at the conclusion of the study will 395 be to examine altered molecular and cellular signatures of neurogenesis related to CN/NFAT treatment 396 that will be valuable to the growing body of research targeting neurogenic mechanisms in AD (Babcock 397 et al., 2021).

398 We did not identify treatment-related modulation to hippocampal volume or to any other brain 399 region by CN inhibition with tacrolimus nor by NFAT inhibition with Q134R. However, our group has 400 previously shown that multi-shell DWI, an imaging technique sensitive to microstructural gray and white 401 matter changes (Alexander et al., 2007; Afzali et al., 2021; Radhakrishnan et al., 2022), showed age-402 related microstructural changes within the hippocampus, parahippocampal gyrus, and prefrontal cortex 403 in placebo-treated dogs that were slower in tacrolimus-treated dogs after just one year (Radhakrishnan 404 et al., 2021). Our negative finding with the current volumetric analysis is not surprising considering that 405 tacrolimus, Q134R and other peptide-based strategies targeting CN/NFAT-related signaling are known to 406 affect cytoarchitectural and biochemical features including synaptic plasticity, neuroinflammation, and glutamate regulation (Dineley et al., 2007; Furman et al., 2012; Hudry et al., 2012; Sompol et al., 2017) 407 408 that macrostructural measurements derived from T1-weighted imaging cannot detect. Furthermore, the 409 majority of the dogs in this study had not yet reached the typical age where the effects of increasing $A\beta$

- 410 burden such as oxidative stress, neurotoxicity, and neuronal loss become more prevalent, which begins
- around nine years old for beagles (developmentally similar to 60-year-old humans). (Cotman & Head,
- 2008). Thus, the final two MRI sessions for the dogs collected at mean ages of 10.5 and 11.5 years will
 offer greater insights into whether chronic CN/NFAT inhibition ameliorates impairments related to Aβ
- 414 burden and/or ultimately slows the trajectory of age-related cortical and subcortical atrophy.
- 415 Importantly, the lack of treatment effects on brain structure and age-associated volumetric losses
- 416 suggests that although there may be no benefits currently on this outcome measure, long-term
- 417 treatment with tacrolimus or Q134R does not appear to accelerate brain atrophy, which was recently
- 418 found to be prevalent across numerous clinical trials for anti-Aβ therapies (Alves et al., 2023). However,
- 419 we will continue to comprehensively monitor the safety and tolerability of these compounds following
- 420 chronic administration in the aging beagles from neuroimaging and other standard safety measures.

421 Across all treatment groups, we observed volume losses across numerous cortical areas as well 422 as ventricular enlargement, all consistent with prior reports of canine neuroimaging findings (Su et al., 423 1998; Kimotsuki et al., 2005). Importantly, our observations recapitulate the differential patterns of 424 volume decline previously observed in aging beagles such as accelerated atrophy of the frontal lobe 425 (Tapp et al., 2006). The prefrontal cortex is an early site of A β accumulation in both aging humans and 426 canines (Head et al., 2000), which makes structural changes to this region an informative neuroimaging 427 biomarker of potential AD-related neurodegeneration. Conversely, its preservation can be used as a 428 biomarker of interest to distinguish effects of CNI or NFAT inhibitor treatment at the conclusion of the 429 study. Interestingly, while a prior cross-sectional study by Tapp et al. (Tapp et al., 2004) suggests frontal 430 lobe atrophy occurs around ten years of age, their work did not include dogs between 7 to 9 years. Our 431 longitudinal findings address this gap and suggest that the onset of frontal lobe volume decline occurs 432 early during middle age, beginning as early as 5 years old in some of the dogs in the present study. 433 Future work in comparison to non-enriched animals can shed light on whether the rate of frontal lobe 434 volume loss may have been attenuated by enrichment.

435 The greater spatial sensitivity of the voxel-based DBM analysis revealed focal areas of significant 436 age-related volume decline at baseline within the cingulate, occipital, and temporal lobes. Among these 437 regions included the posterior cingulate gyrus, a region associated with early A β accumulation 438 (Palmqvist et al., 2017), hypometabolism (Minoshima et al., 1997), synapse loss (Scheff et al., 2015; 439 Scheff & Price, 2001), disrupted functional connectivity (Berron et al., 2020), and atrophy in the earliest 440 phases of AD (Pengas et al., 2010). Prior work in beagles at 14 years old report extensive A β plaque 441 aggregates in the cingulate cortex (Pop et al., 2012), but the time course of neurodegeneration in this 442 region that may precede A β accumulation has not been characterized in beagles during middle age. Our 443 findings provide the first in vivo evidence to our knowledge of posterior cingulate atrophy occurring as 444 early as middle age that parallels human posterior cingulate atrophy and precedes frontal lobe atrophy 445 in canines.

Furthermore, we identified novel volume preservation of the caudate nucleus. Interestingly, our baseline evaluation of cross-sectional volumes with respect to age showed a trend in caudate nucleus volume decline (Fig. 9), suggesting that exposure to our behavioral enrichment protocol may have contributed to its attenuation in volume changes across all groups (Fig. 5a,b). The caudate nucleus is part of an extensive prefrontal-striatal network that is involved in numerous functions, including coordination and motor planning (Fuster, 2002), and its volume steadily declines throughout human
adulthood (Raz et al., 2003). In beagles, the caudate nucleus exhibits reduced glucose metabolism by
middle age (London et al., 1983), and along with the frontal cortex, is at risk for developing lesions in the
form of lacunar infarcts or cysts (Su et al., 2005). More work is necessary to understand whether the

- 455 preserved volumes of the caudate nucleus may be a result of daily exercise and training on numerous
- reward-based visuomotor tasks, similar to preservations previously observed in older human adults
 under long-term motor coordination training (Niemann et al., 2014).
- Additionally, our voxel-wise DBM analysis also uncovered unexpected areas of white matter expansion among areas with well-documented white matter degradation in middle aged humans (Raz et al., 2010). Long-term behavioral interventions such as training on a new visuomotor skill (Scholz et al., 2009), aerobic exercise (Mendez Colmenares et al., 2021), and memory training (De Lange et al., 2018) can induce white matter enhancements in human adults. DWI assessments are currently being performed in the dogs to further characterize these potential enrichment-associated alterations in both gray and white matter microstructure in greater detail.

465 Our ROI-based analyses were important descriptive analyses that can enhance statistical power 466 by limiting the number of tests but can be prone to providing an incomplete picture by potentially 467 artificially segregating the brain. Here, the problem was exacerbated by having several ROIs excluded on 468 the basis of problematic segmentations during image preprocessing. A previous study comparing 469 manual versus automated methods for brain extraction found that manual masking was still superior 470 (Milne et al., 2016), indicating that more work is necessary for refining these image preprocessing 471 procedures for canine data. Recently developed tools have leveraged deep learning for improving the 472 accuracy of brain extraction performance (Tustison et al., 2021; Yu, Han, et al., 2022) but have been 473 trained on human images and have yet to be adapted for canines. However, our novel application of 474 DBM in the canine offers a more agnostic view to volumetric changes with greater spatial precision and 475 was a valuable complementary approach for assessing morphometric alterations in the present study.

We were unable to perform any meaningful evaluations of sex differences because our cohort
of dogs was predominantly female due to the availability of retired female breeders for the study.
However, previous observations of sex differences in canine brain structure (Tapp et al., 2006) as well as
the increasing proportion of women at risk for developing AD dementia over the next four decades
compared to men (Rajan et al., 2021) suggest the need for future investigations to directly assess
potential sex differences in drug efficacy the canine model.

482 Our findings offer novel insights into in vivo volume changes that occur during middle age, prior 483 to the known onset of major A β pathology in aging canines. We demonstrate the feasibility of an 484 automated structural analysis pipeline to assess longitudinal changes in the canine brain at a high level 485 of detail using state-of-the-art neuroimaging analysis tools designed to detect subtle alterations in brain 486 morphology. The application of this analytical framework revealed both well-documented and novel 487 structural alterations to the aging beagle brain and underscore the parallels of cortical and subcortical 488 longitudinal changes in volume between canines and humans during aging. Our future application of this 489 analytical framework will be instrumental for examining the utility of our in vivo neuroimaging 490 biomarkers in predicting disease burden and cognitive outcomes under CN/NFAT inhibition. These 491 findings in a natural model of AD like the beagle offer important contributions to the growing body of

492 493 494	research aimed at understanding the role of modifiable lifestyle factors such as exercise, diet, and cognitive enrichment for reducing the risk of AD (Lista et al., 2015) and suggest that middle age may be a promising therapeutic window of behavioral intervention.
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Group N Age at Start Sex 782 Placebo 14 6.23 ± 0.95 12F, 2M 783 Tacrolimus 15 6.64 ± 1.34 13F, 2M 784 Q134R 14 6.53 ± 1.12 11F, 3M 785 Table 1. Dog characteristics. Ages are presented as mean ± standard deviation (SD). 787	780				
Placebo 14 6.23 ± 0.95 12F, 2M 783 Tacrolimus 15 6.64 ± 1.34 13F, 2M 784 Q134R 14 6.53 ± 1.12 11F, 3M 785 Table 1. Dog characteristics. Ages are presented as mean ± standard deviation (SD). 787	781	Group	Ν	Age at Start	Sex
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Table 1. Dog characteristics. Ages are presented as mean ± standard deviation (SD).	784	Q134R	14	6.53 ± 1.12	11F, 3M
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ROI			Lobe	T0 volume ± SD (mm³)
	1.	Orbital gyrus	Frontal	652.48 ± 70.75
	2.	Pregenual gyrus	Frontal	225.86 ± 21.69
	3.	Gyrus proreus	Frontal	235.91 ± 49.63
	4.	Anterior compositus gyrus	Sensory-motor	727.67 ± 76.88
	5.	Precruciate	Sensory-motor	547.19 ± 68.59
	6.	Genualis	Cingulate	146.08 ± 16.65
	7.	Anterior cingulate gyrus	Cingulate	108.95 ± 9.91
	8.	Medial cingulate gyrus	Cingulate	118.43 ± 11.73
	9.	Posterior cingulate gyrus	Cingulate	375.43 ± 36.94
	10.	Medial occipital gyrus	Occipital	1163.82 ± 108.14
	11.	Recurrens	Occipital	192.04 ± 30.91
	12.	Pararecurrens	Occipital	211.77± 24.82
	13.	Suprasylvian gyrus	Occipital	1285.17 ± 148.69
	14.	Lateral fissure	Occipital	233.51 ± 29.37
	15.	Coronal gyrus	Parietal	230.87 ± 31.41
	16.	Presplenial gyrus	Parietal	170.29 ± 17.94
	17.	Entolateral gyrus	Parietal	459.11 ± 73.22
	18.	Marginal gyrus	Parietal	230.55 ± 22.68
	19.	Posterior compositus gyrus	Temporal	594.14 ± 69.97
. ~	20.	Ectosylvian gyrus	Temporal	1677.71 ± 162.77
	21.	Sylvian gyrus	Temporal	1027.48 ± 108.43
	22.	Amygdala	Subcortical	124.57 ± 11.86
	23.	Caudate nucleus	Subcortical	483.14 ± 44.38
	24.	Hippocampus	Subcortical	520.39 ± 50.12

Table 2. ROIs from the simplified Johnson atlas, lobar locations, and mean volume (mm3) ± SD at the
baseline scan (T0).

Lobe Region Media IQR Medi Ra an Slope n Slope nk IQR Rank 5 Orbital Frontal 0.153 6 952 0.164937 Pregenual 6 Frontal 0.195 8. 518 0.152602 5 Proreus Frontal 0.266 10 5 .5 0.21141 111 Anterior Cingula 0.159 10 8 Cingulate 079 te 0.114573 -0.189 8 9. Precruciate Sensory 43 5 -Motor 0.138192 Genualis Cingula 0.168 17 8. 0.000712 686 5 te Anterior 10 Sensory 0.210 11 Compositus 0.121159 034 -Motor Medial Cingula 0.148 13 10 _ 0.070759 Cingulate te 147 Cingula Posterior 0.196 10 9 te 📍 Cingulate 0.121226 078 Medial Occipit 0.102 14 6 al Occipital 0.077765 80 Recurrens Occipit 0.197 11 13 954 al 0.163141 8 13 Pararecurre Occipit 0.198 721 .5 al 0.117539 ns Suprasylvia Occipit 0.112 17 6. 0.043906 033 5 al n Occipit LateralFissu 0.103 15 9. -0.051082 909 5 re al PRCoronal Parietal 0.091 16 6. -0.01843 906 5

Presplenial		Parietal	-		0.160	12		9.
			0.100011	541			5	
Entolateral		Parietal	0.014		0.177	18		10
			479	597				
PRMarginal		Parietal	-		0.211	11		10
_			0.101112	757			X	
Posterior		Tempor	-		0.152	6	\mathbf{O}	6.
Compositus	al	·	0.180458	34			5	
•							-	
Ectosylvian		Tempor	-		0.097	14		4
	al		0.072358	968				
Sylvian		Tempor	-		0.093	8		5.
	al		0.142836	467			5	
Amygdala		Subcort	-		0.139	17		10
	ical		0.008197	339				
Caudate		Subcort	0.055		0.133	22		4.
Nucleus	ical		629	977			5	
Hippocamp		Subcort	0.157		0.110	24		2
us	ical		849	843				

Table 3. Median volume-over-time slope and ranks for the slopes for each ROI. IQR=inter-quartile range. 794 MeurosciAcu

Region pval_adjusted tstat pval Orbital < 0.0001 < 0.0001 _ 7.787373 Pregenual < 0.0001 < 0.0001 7.875295 Proreus -5.76593 < 0.0001 < 0.0001 AnteriorCompositus -3.49838 0.0011 0.0088 < 0.0001 < 0.0001 Precruciate 6.375448 0.9713 Genualis 0.7728 0.290519 AnteriorCingulate < 0.0001 < 0.0001 -5.889655 MedialCingulate 0.0002 0.0022 4.065612 PosteriorCingulate < 0.0001 < 0.0001 5.035147 MedialOccipital < 0.0001 < 0.0001 5.444237 < 0.0001 < 0.0001 Recurrens 5.300234 < 0.0001 < 0.0001 Pararecurrens -6.048501 Suprasylvian 0.019072 0.9849 0.9849 LateralFissure 0.0006 0.0054 _ 3.689741 PRCoronal 0.5737 0.967 0.567065 Presplenial -4.96738 < 0.0001 < 0.0001 Entolateral 0.768581 0.4464 0.948

PRMarginal	-	0.0084	0.0494	
	2.767238			
PosteriorCompositus	-	< 0.0001	<0.0001	
	8.246205			
Ectosylvian	-	0.0025	0.0174	
	3.211701			X
Sylvian	-	< 0.0001	<0.0001	
	8.558256			
Amygdala	-	0.694	0.9713	
	0.396209		S	
CaudateNucleus	4.093335	0.0002	0.0022	
Hippocampus	7.447361	<0.0001	<0.0001	

797 Table 4. One-sample t-tests against zero for each set of dog-specific volume slopes of each ROI. The

798 Holm-Sidak method was used for adjusting p-values for multiple comparisons.

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0	v	~

	Estimate		С		St		z		Ρ]		0
		oef.		d.Err.				> z		0.025	.975]	
	Intercept		0		0.		0		0	-		0
	·	.088		131		.668		.504		0.17	.345	
	Region[T.Amvgdala]		-		0.		-		0	_	X	0
		0.089		133		0.67		.503	-	0.35	.172	
	Region[T.AnteriorCin		0		0.		0		0	-		0
gulate]		.016		133		.122		.903		0.245	.277	
	Region[T.AnteriorCo		0		0.		0		0			0
mposit	us]	.03		133		.226		.821		0.231	.291	
	Region[T.CaudateNuc		0		0.		0		0	-		0
leus]		.015		133		.112		.911		0.246	.276	
	Region[T.Ectosylvian]		0		0.		0		0	-		0
		.01		133		.075		.94		0.251	.271	
	Region[T.Entolateral]		0		0.	5	0		0	-		0
		.003		133		.02		.984		0.258	.264	
	Region[T.Genualis]		0	0X	0.		0		0	-		0
		.078	C	133	,	.583		.56		0.183	.339	
	Region[T.LateralFissu				0.		-		0	-		0
re]	•	0.114		133		0.853		.394		0.375	.147	
	Region[T.MedialCing		0		0.		0		0	-		0
ulate]	2	.019		133		.139		.889		0.242	.279	
	Region[T.MedialOcci		-		0.		-		0	-		0
pital]		0.071		133		0.536		.592		0.332	.19	
	Region[T.Orbital]		-		0.		-		0	-		0
	6	0.045		133		0.338		.735		0.306	.216	
	Region[T.PRCoronal]		-		0.		-		0	-		0
2		0.01		133		0.075		.94		0.271	.251	
	Region[T.PRMarginal]		0		0.		0		0	-		0
		.018		133		.139		.89		0.242	.279	
	Region[T.Pararecurre		-		0.		-		0	-		0
ns]		0.155		133		1.164		.245		0.416	.106	

Region[T.PosteriorCin		-		0.		-		0	-		0
gulate]	0.029		133		0.216		.829		0.29	.232	
Region[T.PosteriorCo		-		0.		-		0	-		0
mpositus]	0.203		133		1.524		.127		0.464	.058	
Region[T.Precruciate]		-		0.		-		0	-		0
	0.063		133		0.476		.634		0.324	.198	
Region[T.Pregenual]		-		0.		-		0	•-	$\mathbf{\hat{\mathbf{O}}}$	0
	0.04		133		0.3		.764		0.301	.221	
Region[T.Presplenial]		-		0.		-		0			0
	0.012		133	0.	0.087		.93	Ŭ	0.273	.249	Ũ
Degion[T Drorous]				0				0			0
Region[1.Proreus]	0 1 2 8	-	133	0.	0 959	-	338	0	0 389	133	0
	0.120		100		0.555				0.505	.155	-
Region[T.Recurrens]	0 1 9 2	-	122	0.	1 276	-	160	0	-	070	0
	0.183		133		1.376		.169		0.444	.078	
Region[T.Suprasylvia		0		0.	0	0		0	-		0
n]	.013		133	, (.094		.925		0.248	.273	
Region[T.Sylvian]		0		0.		-		0	-		0
			133		0.004		.997		0.261	.26	
BAge_norm		-		0.		-		0	-		0
	0.144	-	092		1.558		.119		0.325	.037	
Time		0		0.		3		<	0		0
	.181	•	048	0.	.77	0	0.001		.087	.275	Ū
Time:Region[T. Amug				0				0			
dalal	0.188	-	068	0.	2.774	-	.006	0	- 0.321	0.055	-
I ime:Region[1.Anteri	0 200	-	068	0.	1 115	-	0 001	<		0 166	-
oreingulatej	0.255		008		4.413		0.001		0.432	0.100	
Time:Region[T.Anteri		-		0.		-		<	-		-
orCompositus]	0.281		068		4.142		0.001		0.414	0.148	
Time:Region[T.Cauda		-		0.		-		0	-		0
teNucleus]	0.132		068		1.954		.051		0.265		
Time:Region[T.Ectosy		-		0.		-		0	-		-
lvian]	0.226		068		3.333		.001		0.359	0.093	
Time:Region[T.Entola		-		0.		-		0			-
teral]	0.163		068		2.404		.016	-	0.296	0.03	

Time:Region[T.Genua	-		0.		-		0	-	-
lis]	0.189	068		2.782		.005		0.321	0.056
Time:Region[T.Latera	-		0.		-		<	-	-
lFissure]	0.248	068		3.666		0.001		0.381	0.116
Time:Region[T.Media	-		0.		-		<	-	-
lCingulate]	0.27	068		3.987		0.001		0.403	0.137
Time:Region[T.Media	-		0.		-		0	-	0 -
lOccipital]	0.231	068		3.41		.001		0.364	0.098
Time:Region[T.Orbita	-		0.		-		<	0	-
l]	0.355	068		5.236		0.001		0.488	0.222
Time:Region[T.PRCor	-		0.		-	Ş	0	-	-
onal]	0.196	068		2.891		.004		0.329	0.063
Time:Region[T.PRMa	-		0.	4	-		<	-	-
rginal]	0.254	068		3.75		0.001		0.387	0.121
Time:Region[T.Parare	-		0.	Ó	-		<	-	-
currens]	0.32	068	. 🤇	4.715		0.001		0.452	0.187
Time:Region[T.Poster	-		0.		-		<	-	-
iorCingulate]	0.295	068		4.353		0.001		0.428	0.162
Time:Region[T.Poster	-		0.		-		<	-	-
iorCompositus]	0.373	068		5.508		0.001		0.506	0.24
Time:Region[T.Precru	-		0.		-		<	-	-
ciate]	0.323	068		4.772		0.001		0.456	0.191
Time:Region[T.Prege	-		0.		-		<	-	-
nual]	0.37	068		5.46		0.001		0.503	0.237
Time:Region[T.Prespl	-		0.		-		<	-	-
enial]	0.269	068		3.972		0.001		0.402	0.136
Time:Region[T.Prore	-		0.		-		<	-	-
us]	0.384	068		5.666		0.001		0.517	0.251
Time:Region[T.Recurr	-		0.		-		<	-	-
ens]	0.327	068		4.826		0.001		0.46	0.194
Time:Region[T.Supras	-		0.		-		0	-	-
ylvian]	0.178	068		2.633		.008		0.311	0.046
Time:Region[T.Sylvia	-		0.		-		<	-	-
n]	0.307	068		4.527		0.001		0.44	0.174

- 801 Table 5. Parameter estimates from the all-ROI-inclusive LME model where the hippocampus was
- 802 assigned as the reference level for the "Region" predictor.

Provide the second seco







Temporal













Metrosci Accepted Martin





Meurosci Accepter

