Basal forebrain activity predicts functional degeneration in the entorhinal cortex and decreases with Alzheimer's Disease progression

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1 Abstract

2	BACKGROUND AND OBJECTIVES Recent models of Alzheimer's Disease (AD) suggest the
3	nucleus basalis of Meynert (NbM) as the origin of structural degeneration followed by the
4	entorhinal cortex (EC). However, the functional properties of NbM and EC regarding amyloid- eta
5	and hyperphosphorylated tau remain unclear.
6	METHODS We analyzed resting-state (rs)fMRI data with CSF assays from the Alzheimer's
7	Disease Neuroimaging Initiative (ADNI, $n=71$) at baseline and two years later.
8	RESULTS At baseline, local activity, as quantified by fractional amplitude of low-frequency
9	fluctuations (fALFF), differentiated between normal and abnormal CSF groups in the NbM but
10	not EC. Further, NbM activity linearly decreased as a function of CSF ratio, resembling the
11	disease status. Finally, NbM activity predicted the annual percentage signal change in EC, but
12	not the reverse, independent from CSF ratio.
13	DISCUSSION Our findings give novel insights into the pathogenesis of AD by showing that local
14	activity in NbM is affected by proteinopathology and predicts functional degeneration within
15	the EC.
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24 Introduction

The basal forebrain's nucleus basalis of Meynert (NbM) has recently been suggested as the 25 26 origin of structural degeneration in Alzheimer's disease (AD) followed by the entorhinal cortex 27 (EC) and other cortical brain regions ^{1,2}. For instance, grey matter loss was more prominent in 28 the NbM compared to the EC in cognitively healthy humans with an abnormal CSF biomarker 29 of amyloid- β (A β) and hyperphosphorylated Tau (pTau). Moreover, the NbM's baseline volume predicted the longitudinal structural degeneration in the EC, further suggesting a trans-synaptic 30 spread of A β starting in the NbM ^{1,2}. This observation in humans is in line with animal work and 31 adds a crucial upstream link to the subsequent spread from EC to other medial temporal lobe 32 33 structures, including the hippocampus, and more distal neocortical brain regions such as the posterior parietal cortex ^{1,3–6}. Importantly, evidence in favor of such a pathological staging 34 35 model is mainly limited to anatomical studies, and, therefore, the functional properties of both the NbM and EC during the disease progression of AD in humans remain unclear. 36

37 Since functional brain changes in AD often precede structural degeneration ^{7–9}, we investigated 38 the functional properties of the NbM and EC, including their functional connectivity. To this end, we used data from the Alzheimer's Disease Neuroimaging Initiative (ADNI) and performed 39 40 a longitudinal region of interest (ROI) analysis over 2 years, focusing on regional and interregional resting-state functional MRI (rsfMRI) properties. In detail, we analyzed (a) the 41 fractional amplitude of low-frequency fluctuations (fALFF) to quantify spontaneous neuronal 42 activity ^{10–12}, (b) regional homogeneity (ReHo) reflecting the synchronicity of neural activity 43 between a voxel and its neighboring voxels ¹³, and finally (c) the functional connectivity 44 between NbM and EC. While all three measures may help to gain new insights into AD 45 progression, we initially focused on fALFF given its established role ^{14–16}, and report ReHo and 46 47 functional connectivity analyses in the supplementary material.

48 In the first step, baseline signals and longitudinal functional changes were compared based on 49 harmonized CSF assays of A β and pTau in NbM and EC. Subsequently, we investigated 50 functional changes in disease progression using the CSF markers. Finally, we tested the 51 competing models NbM \rightarrow EC vs. EC \rightarrow NbM on a functional level. Our main hypothesis was that 52 functional signals in the NbM predict functional change in EC, which would provide further 53 evidence supporting the pathological staging model originating from NbM to EC. From a more 54 general perspective, we aimed to provide new insights into the underlying functional properties 55 of AD, which may contribute to further developing markers and treatment strategies.

56 Methods

57 ADNI data

Data used in the preparation of this article were obtained from the Alzheimer's Disease Neuroimaging Initiative (ADNI) database (adni.loni.usc.edu). ADNI was launched in 2003 as a public-private partnership, led by Principal Investigator Michael W. Weiner, MD, to test whether serial magnetic resonance imaging (MRI), positron emission tomography (PET), other biological markers, and clinical and neuropsychological assessment can be combined to characterize the progression of mild cognitive impairment (MCI) and early Alzheimer's disease (AD).

Since rsfMRI was not acquired in all ADNI cohorts, here data were combined from ADNI-GO, ADNI-2 (ADNI-GO/2) and ADNI 3, downloaded from the Image and Data Archive (IDA) platform run by the Laboratory of Neuro Imaging (LONI) (https://ida.loni.usc.edu). Specifically, we only selected data from participants with CSF biomarkers, and two rsfMRI scans acquired with a delay of two years with the same MR scanner and head coil to ensure within-subject comparability.

71 Image acquisition

72 Participants were scanned at multiple sites equipped with 3-Tesla MRI scanners according to 73 unified ADNI monitoring protocols ¹⁷. To ensure maximum compatibility between the 74 measurements, we followed ADNI's recommendations and included only the basic rsfMRI 75 version but not advanced version of ADNI 3 since it is not compatible with ADNI-GO/2. 76 Moreover, all participants here were examined with the same scanner and head coil for both 77 timepoints, T1 and T2 (https://adni.loni.usc.edu/methods/mri-tool/mri-analysis/). Further, we 78 only included MRI data with excellent, good, or fair quality. For further information on image 79 acquisition, see the supplementary material and http://adni.loni.usc.edu.

80 Data preprocessing

81 Considering their specific scanning parameters such as TR, slice order, and volume number, all data were preprocessed with the Data Processing Assistant for Resting-State fMRI Advanced 82 83 (DPARSFA, http://rfmri.org/dpabi) toolbox version 5 (release 5.2_210501), which is based on 84 the Statistical Parametric Mapping toolbox (SPM 12, https://www.fil.ion.ucl.ac.uk/spm/) for 85 MATLAB[®]. It started with the removal of the first ten volumes and subsequently included the following steps a) slice time correction; b) spatial realignment; c) T1 co-registration to the mean 86 87 functional image; d) CSF, gray and white matter tissue segmentation, and spatial normalization using diffeomorphic anatomical registration using exponential lie algebra (DARTEL) ¹⁸ for T1 88 89 images; e) regression of nuisance variables; f) normalization to MNI space and resampling to 90 an isotropic voxel size of 3 mm of the functional images using the parameters estimated by DARTEL (see supplementary material for a detailed description). 91

92 To reduce the influence of excessive head motion, participants exhibiting more than 3.0 mm of 93 maximum movement and a 3.0-degree rotation angle were discarded. Further, images were 94 visually inspected after co-registration, segmentation, and normalization to guarantee high

quality. This included a specific focus on signal loss and artifacts in our regions of interest (NbM,
EC) by overlaying a ROI mask in standardized space; especially, the EC represents a region that
might often be affected by artifacts ¹⁹. For a detailed description of the preprocessing steps,
excluded participants, ROI definition, and rsfMRI analyses for fALFF, ReHo, and the functional
connectivity, see supplementary material.

100 CSF biomarker

AD neuropathology includes the accumulation of A β resulting in plaques and pTau leading to 101 neurofibrillary tangles ^{20,21}. To better understand how both relate to functional degeneration 102 103 in NbM and EC, we followed previous studies ^{1,2} and used ADNI's CSF samples, produced with a fully automated Elecsys[®] protocol of Aβ and pTau from the first measurement (T1). For each 104 105 participant, we extracted Aβ 1-42 and pTau181 values. Since the protocols by Elecsys[®] are still 106 under development, the results are restricted to a specific technical limit (>1700 pg/mL). Higher values were provided by extrapolation of the calibration curve for research purposes only but 107 108 not diagnostics. Further information on CSF draws and analyses can be found at 109 http://adni.loni.usc.edu.

Here, we analyzed both proteins by using a previously established ratio of pTau / A β , which is known to highly concord with PET measures and clinical diagnoses ^{23,25}. Based on these findings, the standardized and cross-validated cut-off of 0.028 was used to divide the participants into an abnormal (pTau / A $\beta \ge 0.028$) and a normal (pTau / A $\beta < 0.028$) CSF group ^{1,23,25}. Importantly, no participant classified with AD had a normal CSF ratio, but a few (n=10) participants classified with MCI did, which indicates an unclear etiology. Nevertheless, we included them based on biological instead of a syndromal grouping ²⁶.

117 Neuropsychological assessment and clinical diagnosis

All participants underwent a comprehensive neuropsychological test battery. Here, T1 scores 118 are used, including validated memory (MEM) and executive function (EF), based on a 119 120 confirmatory factor analysis ^{27,28}. Memory scores include the AD Assessment Scale, Logical 121 Memory test, Mini-Mental-State Examination (MMSE), Rey Auditory Verbal Learning Test (RAVLT). EF scores are based on the Category Fluency, Digit Span Backwards, Digit Symbol 122 123 Substitution, Trails A and B, and the Clock Drawing tests ^{27,28}. We were also interested in the 124 Montreal-Cognitive-Assessment (MoCA), Sum of Boxes in the Clinical Dementia Rating Scale 125 (CDRSB) and the Alzheimer's Disease Assessment Scale Cognitive (ADAS-Cog 13) to get a deeper 126 understanding of the participants' cognitive profiles (see below).

Furthermore, we included participants' T1 diagnosis made by the ADNI Clinical Core: cognitive
normal (CN) (CDR=0, MMSE=24-30), mild cognitive impairment (MCI) (CDR=0.5, MMSE=24-30),
and Alzheimer's disease (AD) (CDR=0.5-1, MMSE=20-26). These classifications represent widely
used cognitive and functional measures in clinical trials ^{29–31}. Further information regarding
diagnostic is available at http://adni.loni.usc.edu.

132 Participants

133 We included rsfMRI data from ADNI-GO/2 and ADNI 3 – but, importantly, only those that also 134 offered a subject's CSF draw (see below) temporally related to a rsfMRI acquisition (e.g., a participant's screening MRI and baseline lumbar punction measurement). This measurement 135 136 served as T1 measurement in the analyses. To maximize the number of subjects, the second 137 measurement was selected after an interval of 1.5 years ± 12 months (T2)¹. Further details on inclusion and exclusion criteria for participating in ADNI are available under 138 http://adni.loni.usc.edu. In total, 153 participants for ADNI-GO/2 and 141 for ADNI 3 (only basic 139 140 rsfMRI version) fulfilled our inclusion criteria. However, a large proportion had to be excluded

mainly based on fMRI data quality (see supplementary material). Thus, data from n=71 participants were analyzed, which could be further subdivided into those with normal CSF (nCSF, n=37) and abnormal CSF (aCSF, n=34) values (Table 1).

144 Table 1 gives an overview of the participants' demographics, as well as information on APOE4

- genotype and harmonized CSF assay, and Table 2 shows the neuropsychological test results at
- 146 baseline (T1).
- 147 Ethics
- 148 Each center collecting data for ADNI provided an IRB (Institutional Review Board) approval and
- 149 meets ADNI's requirements. Informed consent was obtained from all ADNI participants (for
- 150 more information at http://adni.loni.usc.edu). The analyses presented here were approved by
- 151 the local Ethics Committee of the University of Lübeck and carried out after ADNI's
- recommendations including the approval of the manuscript before submitting to a journal.

153 Data availability statement

154 All data are freely available upon request from the Image and Data Archive (IDA) run by the 155 Laboratory of Neuro Imaging (LONI) (https://ida.loni.usc.edu).

156 Statistical analyses

157 Mixed ANCOVA

Mixed ANCOVAs were carried out for all measures separately (i.e., fALFF, ReHo) to compare baseline signals and the annual percentage signal change (APSC, see below) between regions (NbM and EC as a within-subject factor) and CSF groups (normal and abnormal as a betweensubject factor). Covariates such as age, gender, education, ADNI cohort, and scanner manufacturer were included to adjust for different scan protocols and other potential scannerrelated differences. All 2x2 (region x CSF group) mixed ANCOVAs were carried out in IBM SPSS

statistics version 25 (SPSS) with type III sums of squares, and within-subject effects were
 interpreted without covariates ³².

166 Linear regression of disease status based on CSF marker

To better understand the relationship between disease status and functional MRI properties, CSF ratios (see section CSF biomarker) and functional MRI signals were considered in a linear regression model in SPSS. The functional MRI signal served as dependent variable, and CSF ratio as independent variable. The regression was run with the z-scored data. Subsequently, the dependent overlapping correlations of NbM vs EC with CSF ratio were compared using cocor ^{33,34}.

173 Robust regression

To minimize the influence of outliers, especially in the APSC, robust regression models were carried out in MATLAB® R2020b with fitlm using the bisquare weight function with the default tuning constant. The same covariates as for the mixed ANCOVA were included in the model. Finally, the predictive models (NbM \rightarrow EC and EC \rightarrow NbM) were tested for each CSF group (normal and abnormal) and each functional property (fALFF and ReHo). The data was z-scored before entering the analysis to ensure comparability of the APSC and baseline signal.

180 Moderation analyses of independent samples

181 Moderation analyses were carried out in SPSS using the PROCESS macro ³⁵ for fALFF and ReHo 182 investigating whether CSF group assignment moderates the spread (NbM \rightarrow EC vs. EC \rightarrow NbM) 183 of functional degeneration. Here, CSF group was used as a dichotomous moderator variable. 184 For the construction of products mean-centering was applied, and the heteroscedasticity 185 consistent standard error HC3 (Davidson-MacKinnon) was applied.

187 Annual percentage signal change (APSC)

- 188 The following formula ^{1,36} was used to assess longitudinal APSC in fALFF and ReHo. It accounts
- 189 for the days between both measurements and minimizes the influence of differences between
- 190 both measurements within a subject.

191
$$APSC = \left(\frac{Change \ baseline \ (T2-T1) \ signal}{Baseline \ signal}\right) x \left(\frac{365}{Interscan \ interval \ in \ days}\right) x \ 100$$

192

193 Results

194 CSF grouping strategy and neuropsychological assessments

Based on the CSF grouping strategy, we investigated how aCSF and nCSF groups performed in neuropsychological tests. For each test, one-way fixed effect ANOVAs were carried out with CSF group as factor and age, gender, and education as covariates. As expected, the nCSF group is less affected by cognitive impairment than the aCSF group (see Table 2).

199 Lower fALFF values at baseline in aCSF vs. nCSF in NbM but not EC

200 Baseline fALFF values were compared in NbM and EC further subdivided into CSF groups using 201 a 2x2 mixed ANCOVA. We found a main effect of CSF group (F(1,63)=7.943, p=0.006, partial 202 η^2 =0.112, Fig. 1B), that was driven by lower fALFF values in participants with abnormal CSF, and a significant region x CSF group interaction (F(1,63)=4.623, p=0.035, partial η^2 =0.068, Fig. 1B), 203 204 that was driven by a more pronounced fALFF reduction in the NbM. Post hoc analyses showed 205 that a significant difference in fALFF between nCSF vs aCSF was only observed in NbM 206 (t(69)=3.141, p=0.002) but not EC (t(69)=1.856, p=0.068). There was no main effect of region $(F(1,69)=2.643, p=0.109, partial \eta^2=0.037, Fig. 1B).$ 207

208 Annual percentage signal change in fALFF does not differentiate between CSF groups

209 or regions

220

210 We used a 2x2 mixed ANCOVA to investigate whether the longitudinal indices of APSC in fALFF

of the NbM and EC differentiated between CSF normal vs. abnormal groups. There was no

- significant main effect of CSF group (F(1,63)=2.077, p=0.154, partial η^2 =0.032, Fig. 1C), or
- region (F(1,69)=0.499, p=0.482, partial η^2 =0.007, Fig. 1C), and no significant group x region

214 interaction (F(1,63)=0.367, p=0.547, partial η^2 =0.006, Fig. 1C) in APSC fALFF.

215 NbM's fALFF relates to CSF ratio

216 In a next step, we used linear regressions on fALFF values from NbM and EC, respectively, with

217 CSF ratio as independent variable. It revealed a significant linear effect in the NbM (R²=0.120,

218 F(1, 69)=9.437, p=0.003, Fig. 2A) but not EC (R²=0.031, F(1, 69)=2.206, p=0.142, Fig. 2B). A

219 direct comparison of both correlations (NbM vs EC, one-tailed, which was justified by our a

priori hypotheses) revealed a significant difference that was driven by a more negative

correlation in NbM compared to EC (z=-1,94; p=0.0262; 95% CI: -0.3429 to 0.0015).

222 Baseline signal in NbM predicts annual percentage signal change in fALFF of EC

223 To further address the temporal changes in AD progression, we examined whether the baseline 224 signal in one region predicts the APSC in the other region. Here, in a first step, we used robust regression modeling for both competing models separately for nCSF vs. aCSF. They revealed no 225 226 significant effect for NbM \rightarrow EC in aCSF (R²=0.263, F(7, 26)=1.33, p=0.277, Fig. 3A, Table S1), and 227 no significant effect for NbM \rightarrow EC in nCSF (R²=0.296, F(7, 29)=1.74, p=0.138, Fig. 3B, Table S1). 228 Similarly, there was no significant effect for EC \rightarrow NbM in aCSF (R²=0.137, F(7, 26)=0.587, 229 p=0.76, Fig. 3D, Table S1), and no significant effect for EC \rightarrow NbM in nCSF (R²=0.175, F(7, 29)=0.88, p=0.534, Fig. 3E, Table S1). 230

231 In a second step, we analyzed both groups together by including CSF group in the two 232 competing regression models. Importantly, we observed a statistically significant effect for the 233 model NbM \rightarrow EC (R²=0.235, F(8, 62)=2.39, p=0.026, Fig. 3C, Table S1), with NbM as a significant predictor of EC's APSC (r=-0.3751, t(62)=-3.1445, p=0.003, confidence interval (CI): -0.6136 to 234 0.1366). The other regression model EC \rightarrow NbM did not show a significant effect (R²= 0.0884, 235 236 F(8, 62)= 0.751, p= 0.646, Fig. 3F, Table S1). Replacing CSF as dichotomous predictor by the 237 continuous CSF ratio did not change the results (i.e. significant effects for the model NbM \rightarrow EC, 238 p=0.021, but not EC \rightarrow NbM, p=0.466).

239 CSF group does not moderate the relationship of NbM and EC in fALFF

Finally, we performed two moderation analyses. The first model included baseline fALFF NbM as independent variable, fALFF EC APSC as dependent variable and CSF group as moderator. The model was statistically significant ($R^2=0.2215$, F(9,61)=3.4009, p=0.0019), with a significant direct effect of NbM \rightarrow EC (t(61)=-3.4420, p=0.001), but, no significant moderator effect (t(61)=0.4095, p=0.6836), which is in line with the robust regression analysis.

The second model included baseline fALFF EC as independent variable, fALFF NbM APSC as dependent variable and CSF group as moderator. The model was not statistically significant (R²=0.0965, F(9,61)=0.7857, p=0.6303), which, again, is in line with the robust regression analysis.

249 The results for ReHo and functional connectivity can be found in the supplementary material.

250 Discussion

We investigated the functional properties of the human NbM and EC in relation to the disease progression of AD based on longitudinal rsfMRI data and CSF markers of A β and pTau. With a focus on fALFF, our data provide evidence that spontaneous local brain activity in the NbM, but not EC, is reduced with CSF ratio, and, importantly, it predicts the annual percentage signal change in the interconnected EC independently from proteinopathology. As such, our findings extend previous anatomical studies in humans and animals by providing novel physiological insights into the pathological staging model of AD suggesting the NbM as origin for subsequently affected brain regions possibly via a trans-synaptic mechanism.

259 Local spontaneous brain activity, as quantified by fALFF, was reduced in the NbM at baseline in 260 the abnormal CSF group (Fig. 1B), and there was a linear reduction in fALFF activity with CSF-261 ratio (Fig. 2A). Importantly, both relationships were only observed in the NbM but not in the EC 262 (Fig. 2B), which further underlines that the NbM is specifically vulnerable to AD progression. In fact, pTau and A β are two proteins that have been associated with AD 37 and the NbM is 263 particularly vulnerable to the early accumulation of pTau $^{\rm 38-40}$ and AB deposition $^{\rm 41}.$ This may 264 265 be due to the fact that cholinergic basal forebrain neurons have rather large axons and arbors 266 reaching into the entire central nervous system with high metabolic demands for maintenance, reparation, and transportation ⁴². At the same time, simply due to their sizes, they are more 267 vulnerable to toxins ⁴³, which may further promote disease progression. 268

269 The pathological staging model suggests a structural degeneration originating in the NbM followed by the EC, which adds a crucial upstream link to Alzheimer's degeneration ¹. Our 270 271 functional data support such a view by showing that the NbM's baseline fALFF signal predicted 272 the APSC in the EC (Figure 3C) but not the reverse (Figure 3F). Interestingly, this effect was 273 independent of CSF status, which was further supported by the absence of a moderating effect 274 of CSF. While this is compatible with a specific spread from NbM to EC, it also indicates that the 275 putative functional consequences, namely changes in neural activation, are unrelated to pTau 276 and A β . This apparently differs from anatomical changes from NbM to EC that were more pronounced in subjects with abnormal CSF¹. From a physiological point of view, a trans-277

278 synaptic spread of proteins between anatomically interconnected brain regions is possible and 279 has been shown in several animal studies. For instance, aggregates of tau can propagate from 280 the EC to other limbic regions, including the dentate gyrus and hippocampal CA fields, followed by neocortical brain regions including the parietal cortex ^{3–5,44}. In vitro, this can be enhanced by 281 282 neural activity ⁶, which might help to explain why CSF status did not moderate the relationship 283 between NbM activity and longitudinal changes in EC activity in our study. While this needs to 284 be further investigated using larger and independent samples, our study is the first to show in vivo in humans that a neural signal in NbM can serve as a predictive marker for functional 285 286 changes in the anatomically interconnected EC across healthy controls, MCIs and AD patients. Although fALFF is a prominent marker of spontaneous local brain activity ^{11,12}, only a limited 287 number of studies used fALFF to investigate AD. Importantly, previous work did not specifically 288 289 focus on the NbM and EC but other, typically larger brain regions. It showed, for instance, 290 decreased fALFF signals in the bilateral middle frontal and left precuneus in participants with 291 positive A β^{14} . In preclinical AD, increases and decreases in fALFF were reported in the right 292 inferior frontal gyrus ^{14,45}, and in prodromal AD lower fALFF signals could be shown in the 293 bilateral precuneus, right middle frontal gyrus, right precentral gyrus, and postcentral gyrus. Finally, in AD fALFF was increased in the right fusiform gyrus, medial temporal lobe and inferior 294 295 temporal gyrus, but decreased in the bilateral precuneus, left posterior cingulate cortex, left cuneus and superior occipital gyrus ⁴⁵. These partly divergent effects of fALFF associated with 296 297 AD might be explained by compensatory effects to maintain an adequate level of cognitive 298 performance ⁴⁵, and could be a functional hallmark of neural aging ⁴⁶ that needs further 299 attention. Furthermore, since no significant effects in ReHo and functional connectivity 300 between were detected (see supplementary material), fALFF seems to be a particularly 301 sensitive marker. Together, fALFF is highly sensitive to changes in neural activity associated with
 302 AD even in rather small brain regions and therefore offers a useful marker in future studies.

303 Our analyses specifically focused on the functional properties of the human NbM and EC but 304 no other interconnected brain regions that, according to the pathological staging model, follow 305 the EC. These may include the parahippocampal cortex and hippocampal structures, as well as 306 the parietal cortex ^{1,3–5,44}. Along these lines, we included functional signals averaged from both 307 hemispheres, which simplified our analyses, but it neglected possible lateralization effects ^{47,48}. 308 Second, ADNI is a large multicenter study offering a rich and unique dataset. However, our 309 rsfMRI data come from different MR scanners, possibly leading to a bias in image quality and 310 extracted signal. Therefore, we only included high-quality data that were based on comparable 311 protocols and within-subject measurements from the same scanner. We also employed 312 appropriate covariates in our statistical models, and differences in scanning parameters (e.g. slice order or number of volumes) were accounted for by during preprocessing ^{49,50}. Further, 313 314 our main findings are based on analyses including a measure of APSC, which is robust against 315 within-subject variability, e.g., because of the MR scanner.

316 Functional activity in the human basal forebrain decreased with proteinopathology and

317 predicted the functional decrease within the interconnected EC independent from CSF status.

As such, our findings extend the pathological staging model of AD by giving novel insights into

- 319 the functional properties of the underlying brain regions. From a more general perspective,
- 320 fALFF appears to be a suitable marker to further investigate functional brain changes
- 321 associated with the progression of AD.

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ACKNOWLEDGEMENT LIST FOR ADNI PUBLICATIONS

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363 References

364	1.	Fernández-Cabello S, Kronbichler M, Van Dijk KRA, et al. Basal forebrain volume reliably
365		predicts the cortical spread of Alzheimer's degeneration. Brain 2020;143(3):993–1009.
366	2.	Schmitz TW, Spreng RN. Basal forebrain degeneration precedes and predicts the cortical
367		spread of Alzheimer's pathology. Nat Commun 2016;7(1):13249.
368	3.	de Calignon A, Polydoro M, Suárez-Calvet M, et al. Propagation of tau pathology in a model
369		of early Alzheimer's disease. Neuron 2012;73(4):685–697.
370	4.	Liu L, Drouet V, Wu JW, et al. Trans-Synaptic Spread of Tau Pathology In Vivo. PLOS ONE
371		2012;7(2):e31302.
372	5.	Khan UA, Liu L, Provenzano FA, et al. Molecular drivers and cortical spread of lateral
373		entorhinal cortex dysfunction in preclinical Alzheimer's disease. Nat Neurosci
374		2014;17(2):304–311.
375	6.	Wu JW, Hussaini SA, Bastille IM, et al. Neuronal activity enhances tau propagation and tau
376		pathology in vivo. Nat Neurosci 2016;19(8):1085–1092.
377	7.	Warren SL, Moustafa AA. Functional magnetic resonance imaging, deep learning, and
378		Alzheimer's disease: A systematic review. Journal of Neuroimaging 2023;33(1):5–18.
379	8.	Sperling R. The potential of functional MRI as a biomarker in early Alzheimer's disease.
380		Neurobiology of Aging 2011;32:S37–S43.
381	9.	Johnson KA, Fox NC, Sperling RA, Klunk WE. Brain Imaging in Alzheimer Disease. Cold
382		Spring Harbor Perspectives in Medicine 2012;2(4):a006213–a006213.

- Biswal B, Zerrin Yetkin F, Haughton VM, Hyde JS. Functional connectivity in the motor
 cortex of resting human brain using echo-planar mri. Magn. Reson. Med. 1995;34(4):537–
 541.
- 386 11. Zou Q-H, Zhu C-Z, Yang Y, et al. An improved approach to detection of amplitude of low387 frequency fluctuation (ALFF) for resting-state fMRI: Fractional ALFF. Journal of
 388 Neuroscience Methods 2008;172(1):137–141.
- 389 12. Zuo X-N, Di Martino A, Kelly C, et al. The oscillating brain: Complex and reliable.
 390 NeuroImage 2010;49(2):1432–1445.
- 391 13. Zang Y, Jiang T, Lu Y, et al. Regional homogeneity approach to fMRI data analysis.
 392 NeuroImage 2004;22(1):394–400.
- Wang S-M, Kim N-Y, Kang DW, et al. A Comparative Study on the Predictive Value of
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 Alzheimer's Disease. Front. Psychiatry 2021;12:626332.
- 396 15. Yang L, Yan Y, Wang Y, et al. Gradual Disturbances of the Amplitude of Low-Frequency
 397 Fluctuations (ALFF) and Fractional ALFF in Alzheimer Spectrum. Front. Neurosci.
 398 2018;12:975.
- 399 16. Zhang X, Xue C, Cao X, et al. Altered Patterns of Amplitude of Low-Frequency Fluctuations
 and Fractional Amplitude of Low-Frequency Fluctuations Between Amnestic and Vascular
 401 Mild Cognitive Impairment: An ALE-Based Comparative Meta-Analysis. Front. Aging
 402 Neurosci. 2021;13:711023.

- 403 17. Jack CR, Bernstein MA, Borowski BJ, et al. Update on the Magnetic Resonance Imaging
 404 core of the Alzheimer's Disease Neuroimaging Initiative. Alzheimer's & amp; Dementia
 405 2010;6(3):212–220.
- 406 18. Ashburner J. A fast diffeomorphic image registration algorithm. NeuroImage
 407 2007;38(1):95–113.
- 408 19. Olman CA, Davachi L, Inati S. Distortion and Signal Loss in Medial Temporal Lobe. PLoS
 409 ONE 2009;4(12):e8160.
- 410 20. Palmqvist S, Mattsson N, Hansson O, for the Alzheimer's Disease Neuroimaging Initiative.

411 Cerebrospinal fluid analysis detects cerebral amyloid-β accumulation earlier than positron
 412 emission tomography. Brain 2016;139(4):1226–1236.

- 413 21. Shaw LM, Vanderstichele H, Knapik-Czajka M, et al. Cerebrospinal fluid biomarker
 414 signature in Alzheimer's disease neuroimaging initiative subjects. Ann Neurol.
 415 2009;65(4):403–413.
- Palmqvist S, Zetterberg H, Mattsson N, et al. Detailed comparison of amyloid PET and CSF
 biomarkers for identifying early Alzheimer disease. Neurology 2015;85(14):1240–1249.
- Schindler SE, Gray JD, Gordon BA, et al. Cerebrospinal fluid biomarkers measured by
 Elecsys assays compared to amyloid imaging. Alzheimer's & Dementia 2018;14(11):1460–
 1469.
- 421 24. Witte MM, Foster NL, Fleisher AS, et al. Clinical use of amyloid-positron emission
 422 tomography neuroimaging: Practical and bioethical considerations. Alzheimer's &
 423 Dementia: Diagnosis, Assessment & Disease Monitoring 2015;1(3):358–367.

424	25.	Hansson O, Seibyl J, Stomrud E, et al. CSF biomarkers of Alzheimer's disease concord with
425		amyloid- β PET and predict clinical progression: A study of fully automated immunoassays
426		in BioFINDER and ADNI cohorts. Alzheimer's & Dementia 2018;14(11):1470–1481.
427	26.	Jack CR, Bennett DA, Blennow K, et al. NIA-AA Research Framework: Toward a biological
428		definition of Alzheimer's disease. Alzheimer's & amp; Dementia 2018;14(4):535–562.
429	27.	Crane PK, Carle A, Gibbons LE, et al. Development and assessment of a composite score
430		for memory in the Alzheimer's Disease Neuroimaging Initiative (ADNI). Brain Imaging and
431		Behavior 2012;6(4):502–516.
432	28.	Gibbons LE, Carle AC, Mackin RS, et al. A composite score for executive functioning,
433		validated in Alzheimer's Disease Neuroimaging Initiative (ADNI) participants with baseline
434		mild cognitive impairment. Brain Imaging and Behavior 2012;6(4):517–527.
435	29.	Aisen PS, Petersen RC, Donohue MC, et al. Clinical core of the Alzheimer's disease
436		neuroimaging initiative: Progress and plans. Alzheimer's & Dementia 2010;6(3):239–
437		246.
438	30.	Aisen PS, Petersen RC, Donohue M, et al. Alzheimer's Disease Neuroimaging Initiative 2
439		Clinical Core: Progress and plans. Alzheimer's & amp; Dementia 2015;11(7):734–739.
440	31.	Weiner MW, Veitch DP, Aisen PS, et al. The Alzheimer's Disease Neuroimaging Initiative
441		3: Continued innovation for clinical trial improvement. Alzheimer's & Dementia

442 2017;13(5):561–571.

32. Schneider BA, Avivi-Reich M, Mozuraitis M. A cautionary note on the use of the Analysis
of Covariance (ANCOVA) in classification designs with and without within-subject factors

- 445 [Internet]. Front. Psychol. 2015;6[cited 2021 Oct 6] Available from:
 446 http://journal.frontiersin.org/article/10.3389/fpsyg.2015.00474/abstract
- 33. Diedenhof B, Musch J. Correction: cocor: A Comprehensive Solution for the Statistical
 Comparison of Correlations. PLoS ONE 2015;10(6):e0131499.
- 34. Silver NC, Hittner JB, May K. Testing Dependent Correlations With Nonoverlapping
 Variables: A Monte Carlo Simulation. The Journal of Experimental Education
 2004;73(1):53–69.
- 452 35. Hayes AF. Introduction to mediation, moderation, and conditional process analysis: A
 453 regression-based approach. Guilford publications; 2017.
- 454 36. Cavedo E, Grothe MJ, Colliot O, et al. Reduced basal forebrain atrophy progression in a
- 455 randomized Donepezil trial in prodromal Alzheimer's disease. Sci Rep 2017;7(1):11706.
- 456 37. Alzheimer's Association. Alzheimer's disease facts and figures. Alzheimer's & amp;
 457 Dementia 2021;17(3):327–406.
- 458 38. Mesulam M, Shaw P, Mash D, Weintraub S. Cholinergic nucleus basalis tauopathy emerges
 459 early in the aging-MCI-AD continuum. Annals of Neurology 2004;55(6):815–828.
- 39. Braak H, Del Tredici K. The pathological process underlying Alzheimer's disease in
 individuals under thirty. Acta Neuropathol 2011;121(2):171–181.
- 462 40. Braak H, Del Tredici K. The preclinical phase of the pathological process underlying
 463 sporadic Alzheimer's disease. Brain 2015;138(10):2814–2833.

- 464 41. Baker-Nigh A, Vahedi S, Davis EG, et al. Neuronal amyloid-β accumulation within
 465 cholinergic basal forebrain in ageing and Alzheimer's disease. Brain 2015;138(6):1722–
 466 1737.
- 467 42. Wu H, Williams J, Nathans J. Complete morphologies of basal forebrain cholinergic
 468 neurons in the mouse. eLife 2014;3:e02444.
- 469 43. Mattson MP, Magnus T. Ageing and neuronal vulnerability. Nat Rev Neurosci
 470 2006;7(4):278–294.
- 471 44. Walker LC, Diamond MI, Duff KE, Hyman BT. Mechanisms of Protein Seeding in
 472 Neurodegenerative Diseases. JAMA Neurol 2013;70(3):10.1001/jamaneurol.2013.1453.
- 473 45. Zeng Q, Luo X, Li K, et al. Distinct Spontaneous Brain Activity Patterns in Different
 474 Biologically-Defined Alzheimer's Disease Cognitive Stage: A Preliminary Study. Front.
 475 Aging Neurosci. 2019;11:350.
- 476 46. Cabeza R, Albert M, Belleville S, et al. Maintenance, reserve and compensation: the
 477 cognitive neuroscience of healthy ageing. Nature Reviews Neuroscience 2018;1.
- 478 47. Banks SJ, Zhuang X, Bayram E, et al. Default Mode Network Lateralization and Memory in
 479 Healthy Aging and Alzheimer's Disease. JAD 2018;66(3):1223–1234.
- 480 48. Liu H, Zhang L, Xi Q, et al. Changes in Brain Lateralization in Patients with Mild Cognitive
- 481 Impairment and Alzheimer's Disease: A Resting-State Functional Magnetic Resonance
- 482 Study from Alzheimer's Disease Neuroimaging Initiative. Front. Neurol. 2018;9:3.

483	49.	Badhwar A, Collin-Verreault Y, Orban P, et al. Multivariate consistency of resting-state
484		fMRI connectivity maps acquired on a single individual over 2.5 years, 13 sites and 3
485		vendors. NeuroImage 2020;205:116210.
486	50.	Teipel SJ, Wohlert A, Metzger C, et al. Multicenter stability of resting state fMRI in the
487		detection of Alzheimer's disease and amnestic MCI. NeuroImage: Clinical 2017;14:183-
488		194.
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- 507 Tables
- 508 **Table 1.** Participants' demographics and information on APOE4 genotype and harmonized CSF
- 509 assays

	Normal CSF	Abnormal CSF	Test - χ²/ t
n (total)=71	37	34	χ ² =0.127, p=0.722 χ ² =8.420, p=0.004**
ADNI-GO/2 (n=44) / 3	17/20	27/7	
(n=27)			
CN (n=32) / MCI (n=28) /	27/10/0	5/18/11	χ ² =28.335, p<0.001***
AD(n=11)			
Manufacturer	24/8/5	28/3/3	χ ² =2.959, p=0.228
Philips (n=52) / Siemens			
(n=11) / GE (n=8)			
Age	70.51 (6.23)	72.71 (7.18)	t=-1.376 p=0.173
Female (n=44) / Male	22/15	22/12	χ ² =0.207, p=0.649
(n=27)			
Education (in years)	16.59 (2.44)	15.91 (2.25)	t=1.222, p=0.226
Interscan interval			
In months	22.03 (5.0)	18.74 (6.9)	t=2.294 p=0.025*
In days	685.35 (151.15)	587.76 (210.43)	t=2.227, p=0.030*
APOE 4 (0/1/2)	28/8/1	5/21/8	χ ² =27.224, p<0.001***
Αβ	1430.64 (521.3)	637.22 (187.1)	t=8.670, p <0.001***
рТаи	18.42 (4.83)	40.81 (17.76)	t=-7.114, p<0.001***

Neuropsychological	CSF groups (mean (SD))		F- value	P- value
testing	Normal Abnormal			
MEM score	0.88 (0.6)	-0.08 (0.97)	F(1,66)= 23.4	<0.001*
EF score	1.02 (0.76)	-0.16 (1.1)	F(1,66)= 23.8	<0.001*
MMSE	29.08 (1.12)	25.82 (3.50)	F(1,66)= 23.35	<0.001*
ADAS-Cog 13	9.8 (4.8)	22.89(14.26)	F(1,66)= 26.82	<0.001*
CDRSB	0.3 (0.55)	2.63 (2.36)	F(1,66)= 34.72	<0.001*
МоСА	25.89 (2.34)	20.91 (5.72)	F(1,66)= 22.5	<0.001*
Clock drawing	4.76 (0.55)	4.03 (1.22)	F(1,66)= 8.73	0.004*

511 Table 2. Neuropsychological test results at baseline, compared by CSF normal vs. abnormal

526 Table captions

Table 1. Participants' demographics and information on APOE4 genotype and harmonized CSF assays

Information of the final sample from ADNI-GO/2 and ADNI-3 grouped by CSF. Means and standard deviation (SD) are represented and the respective t-test or chi-square test to investigate possible group differences. Baseline clinical diagnosis: CN=cognitive normal; MCI=mild cognitive impairment; AD=Alzheimer's Disease. Age and education were assessed in years. APOE4 status: no allel / 1 allel / 2 allels. Aβ=amyloid-β in pg/ml as concentration of the amyloid- β 1-42 peptide. pTau=in pg/ml as CSF concentration of hyperphosphorylated tau.

535 *p<0.05, **p<0.01, ***p<0.001

536

537 Table 2. Neuropsychological test results at baseline, compared by CSF normal vs. abnormal

The mean values with standard deviation (SD) of normal vs. abnormal CSF groups. The abnormal CSF group reflected worse performance in all neuropsychological tests. MEM: memory function score; EF: executive function score; MMSE: Mini-Mental State Examination; ADAS-Cog 13: Alzheimer's Disease Assessment Scale- Cognition Subscale, 13 tasks; CDRSB: Clinical Dementia Rating Scale; MoCA: Montreal-Cognitive-Assessment; Clock drawing: clock

543 drawing test

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544 * significant after Bonferroni correction p < 0.05/n (n = 7 tests).
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550 Figure captions

551	Figure 1. Region of interests, baseline signal and annual percentage signal change for fALFF
552	measures. A) ROIs of NbM (red) and EC (blue) on a coronal slice of a T1-weighted standard brain
553	template. Violin plots representing B) the baseline signals at time point 1 (T1) and the signals
554	at the follow-up measurement (T2). C) shows the annual percentage signal change (APSC) in
555	both regions. The horizontal lines represent medians and dotted lines interquartile ranges.
556	*p<0.01.

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Figure 2. Linear regression for z-scored fALFF signal at baseline in A) NbM and B) EC against the z-scored CSF ratio. For the sake of visualization, the diagnosis groups are plotted in different colors and shapes (blue circle for CN, red square for MCI and green rhombus for AD). A significant linear regression was observed only in the NbM (A) but not EC (B), indicating a region specific decrease in functional activity and proteinopathology.

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Figure 3. Adjusted for covariates response plot for the robust regression models in fALFF A) – C) predictive spread in NbM \rightarrow EC and D) – F) predictive spread in EC \rightarrow NbM. On the x-axis zscores for the baseline signal in the respective region, on the y-axis the z-scored annual percentage signal change (APSC). There was a significant spread in C) NbM \rightarrow EC with CSF included in the model. There was no predictive spread in EC \rightarrow NbM. *p<0.05.





