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NUTRIENT PATTERNS AND BRAIN BIOMARKERS OF ALZHEIMER'S DISEASE IN COGNITIVELY NORMAL INDIVIDUALS

V. BERTI^{1,2}, J. MURRAY¹, M. DAVIES¹, N. SPECTOR¹, W.H. TSUI¹, Y. LI¹, S. WILLIAMS¹, E. PIRRAGLIA¹, S. VALLABHAJOSULA³, P. MCHUGH¹, A. PUPI², M.J. DE LEON¹, and L. MOSCONI¹

¹New York University School of Medicine, New York, NY

²University of Florence, Italy

³Weill Cornell Medical College, New York, NY

Abstract

Objectives—Epidemiological evidence linking diet, one of the most important modifiable lifestyle factors, and risk of Alzheimer's disease (AD) is rapidly increasing. However, there is little or no evidence for a direct association between dietary nutrients and brain biomarkers of AD. This study identifies nutrient patterns associated with major brain AD biomarkers in a cohort of clinically and cognitively normal (NL) individuals at risk for AD.

Design—Cross-sectional study.

Setting—Manhattan (broader area).

Corresponding author: Dr. Lisa Mosconi, Department of Psychiatry, NYU School of Medicine, 145 East 32nd St, 2nd Floor, New York NY, 10016. Tel: (212) 263-3255, Fax: (212) 263-3270, lisa.mosconi@nyumc.org.

Contributions: Dr. Berti – study concept and design, analysis and interpretation, critical revision of the manuscript for important intellectual content; Mr. Murray – acquisition of data, analysis and interpretation, critical revision of the manuscript for important intellectual content; Ms. Davies – acquisition of data, critical revision of the manuscript for important intellectual content; Ms. Spector; Dr. Tsui – acquisition of data, analysis and interpretation, critical revision of the manuscript for important intellectual content; Dr. Li – acquisition of data, analysis and interpretation, critical revision of the manuscript for important intellectual content; Ms. Williams – acquisition of data, analysis and interpretation, study supervision; Dr. Pirraglia – analysis and interpretation, critical revision of the manuscript for important intellectual content; Dr. Vallabhajosula – acquisition of data, analysis and interpretation, critical revision of the manuscript for important intellectual content; Dr. McHugh – study concept and design, acquisition of data, critical revision of the manuscript for important intellectual content; Dr. Pupi – analysis and interpretation, critical revision of the manuscript for important intellectual content; Dr. de Leon – study concept and design, analysis and interpretation, critical revision of the manuscript for important intellectual content; Dr. Mosconi – study concept and design, acquisition of data, analysis and interpretation, critical revision of the manuscript for important intellectual content, study supervision. Statistical Analyses were done by Valentina Berti, Lisa Mosconi and Elizabeth Pirraglia

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Participants—Fifty-two NL individuals (age 54+12 y, 70% women, Clinical Dementia Rating=0, MMSE>27, neuropsychological test performance within norms by age and education) with complete dietary information and cross-sectional, 3D T1-weighted Magnetic Resonance Imaging (MRI; gray matter volumes, GMV, a marker of brain atrophy), 11C-Pittsburgh compound-B (PiB; a marker of fibrillar amyloid- β , A β) and 18F-fluorodeoxyglucose (FDG; a marker of glucose metabolism, METglc) Positron Emission Tomography (PET) scans were examined.

Measurements—Dietary intake of 35 nutrients associated with cognitive function and AD was assessed using the Harvard/Willet Food Frequency Questionnaire. Principal component analysis was used to generate nutrient patterns (NP) from the full nutrient panel. Statistical parametric mapping and voxel based morphometry were used to assess the associations of the identified NPs with AD biomarkers.

Results—None of the participants were diabetics, smokers, or met criteria for obesity. Five NPs were identified: NP1 was characterized by most B-vitamins and several minerals [VitB&Minerals]; NP2 by monounsaturated and polyunsaturated fats, including ω -3 and ω -6 PUFA, and vitamin E [VitE&PUFA]; NP3 by vitamin A, vitamin C, carotenoids and dietary fibers [Anti-oxidants&Fibers]; NP4 by vitamin B12, vitamin D and zinc [VitB12&D]; NP5 by saturated, trans-saturated fats, cholesterol and sodium [Fats]. Voxel-based analysis showed that NP4 scores [VitB12&D] were positively associated with METglc and GMV, and negatively associated with PiB retention in AD-vulnerable regions ($p<0.001$). In addition, both METglc and GMV were positively associated with NP2 scores [VitE&PUFA], and negatively associated with NP5 scores [Fats] ($p<0.001$), and METglc was positively associated with higher NP3 scores [Anti-oxidants&Fibers] ($p<0.001$). Adjusting for age, gender, ethnicity, education, caloric intake, BMI, alcohol consumption, family history and Apolipoprotein E (APOE) status did not attenuate these relationships. The identified ‘AD-protective’ nutrient combination was associated with higher intake of fresh fruit and vegetables, whole grains, fish and low-fat dairies, and lower intake of sweets, fried potatoes, high-fat dairies, processed meat and butter.

Conclusion—Specific dietary NPs are associated with brain biomarkers of AD in NL individuals, suggesting that dietary interventions may play a role in the prevention of AD by modulating AD-risk through its effects on A β and associated neuronal impairment.

Keywords

Alzheimer’s disease; nutrition; aging; Positron Emission Tomography (PET); Magnetic Resonance Imaging (MRI)

Introduction

There is increasing evidence to suggest that diet, one of the most important modifiable lifestyle factors, may play a role in preventing or delaying cognitive decline and Alzheimer’s disease (AD), a major public health problem (1–7). AD is the most common cause of dementia and is associated with presence of amyloid-beta (A β) plaques, neurofibrillary tangles and neuronal loss. As pharmacological treatments for AD are limited, there is a growing interest in understanding how diet could mitigate AD risk and progression (8, 9). Despite studies showing protective effects of several nutrients against AD, the overall

picture remains equivocal (10). These studies would greatly benefit from biomarkers for early AD pathology and associated neuronal injury, which are needed to assess the impact of diet on brain health and to monitor treatment efficacy (10), especially during the recently conceptualized preclinical period of AD (11). In vivo biomarkers are needed to clarify how nutrition promotes healthy brain aging (10), and can therefore be protective against AD, which is critical prior to implementing dietary recommendations for prevention and treatment.

There are very few studies that examined the relationships between dietary nutrients and brain biomarkers of AD in cognitively normal (NL) individuals. A few Magnetic Resonance Imaging (MRI) studies investigated the relationship between ω 3 polyunsaturated fatty acids (PUFA) and brain volumes in non-demented elderly, and showed a correlation between higher baseline ω 3-PUFA levels and lower atrophy rates over time (12–14). However, according to current hypothetical models of AD progression, structural MRI changes are secondary to A β deposition and neuronal hypometabolism (11), and previous studies included only individuals of age >65 y. To our knowledge, there are no published studies that examined the the associations of dietary nutrients with brain Abeta and metabolic activity in NL individuals.

The goal of this study was to examine the relationships between dietary nutrient patterns (NPs) and three major AD-biomarkers: brain A β load (i.e., a hallmark of AD pathology) assessed using 11C-Pittsburgh Compound-B (PiB) Positron Emission Tomography (PET), glucose metabolism (METglc, i.e. a proxy for neuronal activity) assessed using 18F-fluorodeoxyglucose (FDG) PET, and gray matter volumes (GMV, a marker of brain atrophy) on MRI in a cohort of young to late middle aged NL individuals. Using these imaging techniques, several studies have shown preclinical biomarker abnormalities in non-demented individuals several years, if not decades, prior to AD symptoms (11).

Given the interactive nature of nutrient action and metabolism, in this study we used principal component analysis (PCA) to generate NPs from a panel of 35 nutrients which have been related to AD or cognitive function or which are known to interact with those AD-related nutrients. NPs are advantageous as they capture the interactive effect of nutrients in combination (15–17). The present multi-modality brain imaging study uses voxel-based analysis techniques such as Statistical Parametric Mapping and Voxel-Based Morphometry to simultaneously examine A β deposition, METglc and GMV to define which NPs are protective against AD (as reflected in lower brain A β , higher metabolic activity and larger GMV among NL individuals, controlling for AD-risk factors such as age, gender, education, ethnicity, BMI, alcohol consumption, family history of AD, and Apolipoprotein E (APOE) genotype.

Methods

Participants

Among a larger pool of clinically and cognitively normal (NL) individuals participating in longitudinal brain imaging studies at New York University (NYU) Langone School of Medicine, this study focused a sub-set of 65 NL participants who were invited to participate

in a lifestyle survey between 2013–2014. This study examined 52 participants who completed all clinical, MRI, PiB- and FDG-PET exams and dietary questionnaires within 6 months of each other. Of the remaining 13 subjects, 9 did not receive either the PiB or the FDG scan, and 4 returned only partially completed dietary questionnaires and were excluded from this examination. Subjects were derived from multiple community sources, including individuals interested in research participation, family members and caregivers of impaired patients. Informed consent was obtained from all subjects for participation in this NYU institutional review board-approved study.

Individuals with medical conditions or history of conditions that may affect brain structure or function, i.e. stroke, diabetes, head trauma, any neurodegenerative diseases, depression, hydrocephalus, intracranial mass, and infarcts on MRI, and those taking psychoactive medications were excluded. Subjects were 25–72 y of age, with education >12 y, Clinical Dementia Rating (CDR)=0, Global Deterioration Scale (GDS)<2, Mini Mental State Examination (MMSE)>28, Hamilton depression scale<16, Modified Hachinski Ischemia Scale<4 and normal cognitive test performance for age and education (18). A family history of late-onset AD that included at least one 1st degree relative whose AD onset was after age 60 was elicited using standardized questionnaires (18–20). APOE genotypes were determined using Polymerase Chain Reaction (PCR) using standardized protocols (21).

Dietary assessments

Dietary data regarding average food consumption over the prior year were obtained using the 116-item version of Harvard/Willet's semi-quantitative food frequency questionnaire (SFFQ) (22, 23). Trained interviewers administered the SFFQ in English. The SFFQ has been validated for determination of nutrient intake in the elderly and young adults (22, 23) and against plasma measurements (24–26). The validity (using two 7-day food records) and reliability (using two 3-month frequency assessments) of various components of the SFFQ was replicated by several studies (3, 4, 27, 28). The food items were categorized into 30 food groups based on similarities in food and nutrient composition, and intake (g/day) of each food group was calculated by summing the intakes of member food items. The daily intake of nutrients from food sources was computed by multiplying the consumption frequency of each portion of every food by the nutrient content of the specified portion (22). The daily total caloric intake (kilocalories) was included as a confound.

A panel of 35 nutrients that have been associated with cognitive function and AD was examined, including fats: monounsaturated fatty acid (MUFA), ω -3 polyunsaturated fatty acid (PUFA), ω -6 PUFA, other PUFA, saturated fatty acid (SFA), trans-saturated fats and cholesterol [6, 29–34]; vitamins and precursors: α - and β -carotene, β -cryptoxanthin, β - γ - and δ -tocopherol, vitamin A, B vitamins including B1, B2, B3, B6, B9 (folate) and B12, vitamin C, vitamin D, vitamin E, lycopene, lutein and zeaxanthin (7, 27, 35–41); minerals: calcium, copper, iron, magnesium, phosphorus, potassium, selenium, and zinc (8, 42); and dietary fibers (43). As moderate alcohol drinking may be protective against dementia [28], alcohol intake (g/day) was also calculated.

Brain imaging

All subjects received volumetric 1.5 T MRI (124 slice T1-weighted Fast-Gradient-Echo, 1.2 mm sections, no interslice gaps), PiB- and FDG-PET scans following standardized procedures (18–20, 44, 45). For PET, subjects were positioned in the scanner 60 min after injection of 15 mCi of ¹¹C-PiB, and scanned for 30 min in 3D-mode on an LS Discovery or BioGraph PET/CT scanner. The FDG scan was performed 30 min after completion of the PiB scan or on a separate day. After an overnight fast, subjects were injected with 5 mCi of ¹⁸F-FDG, positioned in the scanner 35 min after injection, and scanned for 20 min. All images were corrected for photon attenuation, scatter, and radioactive decay and smoothed for uniform resolution (46).

Image analysis was done blind to clinical data. For each subject, summed PET images corresponding to 40–60 min of FDG data and 60–90 min of PiB data were coregistered to MRI using the Normalized Mutual Information (NMI) routine of Statistical Parametric Mapping (SPM8) (47). Parametric standardized uptake value ratio (SUVR) images were generated by normalizing PiB uptake by cerebellar grey matter uptake (48) and FDG by pons activity (49). MRIs were segmented into grey (GM), white matter (WM) and cerebrospinal fluid (CSF) and normalized to Montreal Neurological Institute (MNI) space by high-dimensional warping (DARTEL) using VBM8 (47). MRI-coregistered PET scans were spatially normalized using subject-specific transformation matrixes obtained from MRI, and smoothed with a 10mm FWHM filter.

MRIs were examined using voxel-based morphometry (VBM) (47, 50). A custom template was created using MRI from all subjects by normalizing and segmenting the MRIs using the unified segmentation model with the MNI template and tissue probability maps (TPMs), and averaging the normalized subject TPMs. Individual scans were then processed using the custom TPMs. Jacobian modulation was applied to restore absolute GM volumes (GMV) in the GM images, which were smoothed with an 8-mm FWHM kernel. Total GM, WM, CSF and intracranial volumes were calculated.

Statistical Analysis

SPSS v.21 (SPSS Inc., 2013) and SPM8 were used for data analysis. Clinical and demographical measures were examined using descriptive statistics.

Nutrient pattern construction—Nutrient patterns (NPs) were derived from the pre-defined panel of 35 nutrients using multivariate analysis (principal component analysis, PCA). The PCA is used to derive factor scores (i.e. patterns), which are linear combinations of all of the variables (i.e., nutrients), and with each factor score being weighted towards groups of variables with the highest association with each other. This process generates principal components, whereby principal components can be characterized into patterns. There are no pre-defined hypotheses as to which nutrient variables will be included in the patterns as the classification of variables into patterns is based on statistical associations between variables. Furthermore, the factor scores are derived to be totally independent from each other (the correlation between them is 0), so that, by using factor scores, one not only reduces the number of variables but also reduces any potential problems with

multicollinearity in the model. With this procedure, the factor scores do not interfere with each other, yielding more confidence in the accuracy of the p-values

Five distinct nutrient patterns (NPs) were extracted from the original set of nutrients via PCA (rotation method: varimax with kaiser normalization). An eigenvalue >1.0 was set a priori to determine the NPs to carry into hypothesis testing. Each participant receives a standardized NP score for each pattern that corresponds to a linear combination of the nutrients that load heavily within each pattern.

Primary models—Linear regressions were used to test for associations between demographical and neuropsychological measures (dependent variables) and each NP (independent variables) at $p < 0.05$. The General Linear Model (GLM, i.e. multiple regressions) with post-hoc linear t-contrasts implemented in SPM8 was used to test for associations between FDG, PiB and MRI scans (dependent variables), each NP (independent variables), and confounds. Positive and negative associations between biomarkers and NPs were examined. Age and total caloric intake were included as covariates in all analysis. Gender, education, ethnicity, body mass index (BMI), alcohol consumption, APOE and family history were then examined as covariates in separate models to avoid over-fitting. Education and BMI were modeled as continuous variables. Gender (male vs. female), FH (positive vs. negative, FH+ vs. FH-), and APOE status (APOE4 carriers vs. non-carriers, APOE4+ vs. APOE4-) were examined as dichotomous variables. Ethnic group was based on self-report using the format of the 1990 census. Ethnicity was used as a dummy variable (White/non-Hispanic vs. other ethnic groups). Alcohol intake was used as a dichotomous variable (mild-moderate (0–30 g/day) vs. no (0 g/day) or more than moderate (>30 g/day) consumption) (16). No proportional scaling was performed as PET scans were normalized to reference regions. GMV images were corrected for total intracranial volume.

As we had specific a priori hypotheses on which brain regions would show possible biomarker effects, results were examined at $p < 0.001$, uncorrected (cluster extent >20 voxels, i.e., >2 times the FWHM) (47), in the search volume defined by a masking image created from a set of predefined AD-vulnerable regions, which included: posterior cingulate cortex/precuneus, inferior and superior parietal lobule, lateral and medial temporal cortex (including hippocampus, amygdala and parahippocampal gyrus), medial and prefrontal cortex, striatum (18–20, 44, 45). Anatomical location of brain regions showing significant effects was described using Talairach and Tournoux coordinates, after conversion from MNI space.

Food sources—The food sources of the NPs were examined by testing for correlations between foods and NPs using Pearson's coefficients of determination. Results were considered significant at $p < 0.05$ (2-sided tests).

Results

Subjects' characteristics are found in Table 1. None of the participants were diabetics, smokers, or met criteria for obesity as defined by a Body-Mass index (BMI) >30 kg/m².

Nutrient patterns

Table 2 displays the composition of 5 extracted NPs (all loading coefficients >0.55), which accounted for 86% of the total variance in the nutrient panel. NP1 was characterized by B-vitamins (B1, B2, B3, B6 and B9) and several minerals (i.e., calcium, iron, magnesium, phosphorus, potassium and selenium) [VitB&Minerals]; NP2 by monounsaturated and polyunsaturated fats, including ω -3 and ω -6 PUFA, and vitamin E (mostly β - and γ -tocopherol) [VitE&PUFA]; NP3 by vitamin A, carotenoids (α - and β -carotene, β -cryptoxanthin, lutein and zeaxanthin), vitamin C, and dietary fibers [Anti-oxidants&Fibers]; NP4 by vitamin B12, vitamin D and zinc [VitB12&D]; NP5 by saturated, trans-saturated fats, cholesterol and sodium [Fats].

Nutrient patterns and clinical-demographic characteristics—As shown in eTable 1, age, gender, education, ethnicity, FH, APOE4 carrier status and BMI were not associated with any NP. Higher NP2 scores [VitE&PUFA] were associated with a more favorable HDL/LDL and hip-to-waist ratio ($p<0.05$). Systolic blood pressure was negatively associated with NP1 [VitB&Minerals] and positively associated with NP5 scores [Fats] ($p<0.05$). NP5 scores [Fats] were also positively associated with plasma triglycerides ($p<0.05$) and showed a trend for an association with BMI ($p=0.08$) and homocysteine levels ($p=0.09$). There were no associations between NPs and neuropsychological test scores.

Nutrient patterns and AD-biomarkers

FDG-PET—Results from voxel-based analysis are reported in terms of statistical parametric maps (SPMs), e.g. brain areas showing significant associations between NPs and brain biomarkers. As shown in Table 3, SPMs are defined in terms of cluster extent (i.e., number of voxels in the cluster), cluster coordinates in the Talairach space (x, y, z), Z and P values for each set of coordinates, anatomical description of the brain regions included in each cluster, and corresponding Brodmann areas. Correcting for age and total caloric intake, NP2 scores [VitE&PUFA] were positively associated with METglc in medial, inferior and lateral frontal cortex, bilaterally ($p<0.001$, Table 3). NP3 scores [Anti-oxidants&Fibers] were positively associated with METglc in middle frontal and cingulate cortex of the left hemisphere ($p<0.001$, Table 3). NP4 scores [VitB12&D] were positively associated with METglc in several temporal regions, including superior and medial temporal areas, bilaterally ($p<0.001$, Table 3). NP5 scores [Fats] were negatively associated with METglc in middle and inferior temporal cortex, bilaterally, right frontal cortex, and left parietal cortex ($p<0.001$, Table 3). Adjustment for gender, education, ethnical group, BMI, APOE, family history, and alcohol consumption did not attenuate these relationships (Figure 1). There were no other brain regions showing positive or negative correlations with the remaining NPs.

MRI—NP4 scores [VitB12&D] were positively associated with GMV in temporal and frontal cortex, mostly of the right hemisphere ($p<0.001$, Table 4). NP5 scores [Fats] were negatively associated with GMV in frontal cortex ($p<0.001$, Table 4). Adjustment for gender, education, ethnical group, BMI, APOE, family history, and alcohol consumption did not attenuate these relationships (Figure 1). There were no other brain regions showing positive or negative correlations with the remaining NPs. Overall, with regard to the number

of regions affected and magnitude of impairment in those regions, the association between NPs and biomarkers were less robust with MRI than with FDG measures (Figure 1).

PiB-PET—Only NP4 scores [VitB12&D] were negatively associated with PiB retention in parietal, frontal and PCC regions ($p<0.001$, Table 5). Adjustment for gender, education, ethnical group, BMI, APOE, family history, and alcohol consumption did not attenuate these relationships (Figure 2). There were no other brain regions showing positive or negative correlations with the remaining NPs.

Food sources

Correlations between NPs and food groups showed that NP1 [VitB&Minerals] was mainly from low fat dairies, whole grains and cereals, fresh fruit, nuts, green leafy and cruciferous vegetables with correlation coefficients (r) of 0.47, 0.40, 0.37, 0.33, 0.29 and 0.27 respectively ($p's<0.05$); NP2 [VitE&PUFA] was from vegetable oil, nuts, fish, green leafy and other vegetables, and fresh fruit ($r = 0.61, 0.57, 0.36, 0.32$ and 0.32 ; $p's<0.05$); NP3 [Anti-oxidants&Fibers] was from fresh fruit, green leafy cruciferous and dark leafy vegetables, legumes, and fruit juice ($r = 0.63, 0.63, 0.62, 0.58, 0.47$ and 0.34 , $p's<0.02$); NP4 [VitB12&D] was from fish, eggs, tomato, and low-fat dairies ($r = 0.63, 0.44, 0.30$ and 0.28 , $p<0.05$); and NP5 [Fats] was from sweets, fried potatoes, high-fat dairies, processed meat and butter ($r = 0.45, 0.32, 0.30, 0.28$ and 0.27 , $p's<0.05$).

Discussion

Present results show an association between nutrient patterns, as derived from principal component analysis, and three major brain AD-biomarkers in NL individuals. Specifically, the pattern characterized by vitamin B12, vitamin D and zinc (NP4) was significantly associated with all biomarkers, so that the higher intake of these nutrients, the lower A β load, and the higher METglc and GMV in AD-vulnerable regions. Additionally, metabolic activity and GMV were negatively associated with intake of saturated, trans-saturated fats, cholesterol and sodium (NP5). Finally, METglc was positively associated with intake of vitamin E, mono- and polyunsaturated fats such as $\omega 3$ and $\omega 6$ -PUFA (NP2), and with carotenoids, vitamin A, vitamin C and dietary fibers (NP3). Results were independent of AD-risk factors such as age, gender, education, ethnicity, FH and APOE status, as well as BMI and alcohol consumption. The identified 'AD-protective' patterns were linked to higher intake of vegetables, fruit, whole grains, fish, low fat dairies and nuts, and lower intake of sweets, fried potatoes, processed meat, high-fat dairies and butter, indicating that such dietary pattern might be particularly useful to support brain aging.

Most studies on nutrition and risk for AD investigated nutrients in their isolated forms, although the human diet is strongly influenced by synergy or antagonism among components (17). In this study, we used PCA to generate NPs from a panel of nutrients that have been related to AD or cognitive function, and their combinations. The nutrient pattern associated with all biomarkers (NP4) included neuro-protective vitamin B12 (39, 51), vitamin D (40, 41) and zinc, one of the most important transitional metals for human metabolism that is involved in A β adhesiveness and amyloid precursor protein synthesis (52). A recent randomized controlled trial showed that treatment with a combination of B

vitamins, but especially B12, slowed gray matter volume loss in AD-regions of amnesic MCI patients (53). Our findings suggest that vitamin B12 may have beneficial effects on AD-biomarkers also during the normal stages of cognition.

Higher intake of fats (NP5) was associated with reduced METglc and GMV, consistent with the notion that “bad fats” may have negative effects on cognitive function (31, 32). Additionally, METglc was positively associated with higher intake of vitamin E, a strong anti-oxidant (35, 38), and fatty acids such as ω -3 and ω -6PUFA which are known for their neuroprotective properties through anti-inflammatory, antioxidant, and energy metabolism pathways (29). These results are consistent with previous MRI studies of NL elderly that showed a correlation between higher ω 3-PUFA levels and lower atrophy rates of the medial temporal lobes over time (12–14). Together with previous studies, our data from a substantially younger population (mean age 54 y vs. 72–78 y (12–14)) free of possible confounding comorbidities such as depression or diabetes (12–14), indicate that AD biomarkers are affected by dietary food intake already at middle age, further supporting an early role for nutrition in the prevention of AD. METglc was also positively associated with intake of vitamin A, C, several carotenes, and dietary fibers. These nutrients are known to have beneficial effects via their strong antioxidant and A β anti-oligomerization effects (7, 35, 37, 38, 54), and dietary fibers help regulate glucose levels (43).

A community-based study of NL elderly reported an association between a diet rich in ω -3PUFA and lower plasma A β levels (30). Our results showed an association between higher PUFA/vitamin E scores (NP3), higher metabolic activity and GMV, as well as with a more favorable HDL/LDL ratio, though there were no significant associations with PiB retention. Differences may depend on measurement of central nervous system vs peripheral (plasma) A β measures, as well as on age effects, as our NL cohort was substantially younger than in the previous report (mean age 75 y (30)) and fibrillar A β accumulation increases in an age-dependent fashion (55). Nonetheless, it is possible that FDG and MRI effects reflect toxic effects of A β oligomers, which are known to occur early in AD (55) and which are not currently visible on PET imaging. It is also possible that the combination of PUFA with other nutrients into a single pattern might overshadow the specific impact of PUFA on PiB retention. To our knowledge there are no previous studies that examined the correlation between ω -3PUFA and PiB-PET in NL individuals. While the present study aimed at identifying patterns of nutrients rather than individual nutrients, other studies are warranted to test for associations between brain A β and specific nutrients such as ω -3PUFA.

NP1 [VitB&Minerals] was the only pattern that did not show significant associations with AD-biomarkers. This pattern included several B vitamins and minerals which have been related to better cognitive functioning or lower AD risk in the elderly (27, 36, 39, 42, 51). While all these nutrients loaded into the same PCA-derived NPs, in keeping with their interactive biochemical qualities, their data reduction into a single pattern might hinder detection of each nutrient’s specific effect on biomarkers. To our knowledge, there are no previous studies that investigated the associations of individual B vitamins and minerals with brain AD biomarkers in NL individuals. More studies with larger samples and longitudinal examinations are needed to replicate these preliminary results and investigate whether isolated nutrients show effects.

The food sources associated with our biomarker-identified NPs are consistent with current definitions of an AD-preventative diet, in terms of recommended food choices (for review see (56)). Present data are consistent with epidemiological findings showing that high adherence to dietary patterns characterized by higher intakes of fruits, vegetables, fish, nuts and legumes, and lower intake of meat, high-fat dairies and sweets, is consistently associated with reduced risk for AD (3, 4, 16, 17, 57–59). Our biomarker data supports previous epidemiological studies by offering a possible pathophysiological substrate to the observed reduced risk for AD. The NPs identified using brain imaging in the present study provide biological evidence that a healthy diet is associated with a more favorable “AD-brain profile”, characterized by lower A β load, higher metabolic activity and GMV in NL individuals, several years prior to possible symptoms onset.

Several data reduction techniques have been proposed for analysis of nutrients, including PCA (17). Bowman et al. (15) applied PCA to analysis of plasma nutrients and showed that a nutrient pattern characterized by higher intake of anti-oxidants, vitamin B12, vitamin D and ω -3PUFA was associated with more favorable cognitive and MRI white matter hyper-intensity profiles in NL elderly (15). More studies are needed to assess the relationships between brain imaging of AD pathology and NPs, from both SFFQs and plasma measures. While SFFQs are fairly comprehensive and SFFQ-derived dietary patterns remain quite stable over time (16, 23, 60), this method may be subject to faulty recall of dietary intake and portion size. In support of the accuracy of SFFQ, we detected the expected associations between reported diet and objective measures, e.g. higher NP2 [VitE&PUFA] scores were associated with a more favorable HDL/LDL ratio; higher NP5 [Fats] scores were positively associated with higher systolic blood pressure, plasma triglycerides, homocysteine and BMI. Moreover, the SFFQ used in our study was validated against plasma nutrient measures by several investigators, showing good correlations (24–26).

In our data set, none of the NPs were associated with neuropsychological measures, possibly because our subjects were relatively young to late-middle aged adults and all high-school graduates, which resulted in a “ceiling-effect”. Longitudinal studies with larger samples are warranted to assess whether the relationship between nutrients, AD-biomarkers and cognitive performance varies with age and disease. There is evidence that dietary interventions can alleviate AD pathology in MCI patients (61), consistent with our observations that diet might modulate AD-risk through its effects on A β and associated neuronal dysfunction.

This study has several limitations. The NL population selected in our study consists of a group with a high a priori risk of preclinical AD-changes, and results were found in small numbers of carefully screened subjects under controlled clinical conditions. As we did not request participants to modify their diet in any way, some were taking supplements at the time of the exam. Descriptively, 31/52 (60%) participants reported taking no supplements for >1 year prior to brain imaging, and the remaining 21/52 (40%) subjects reported taking a multivitamin regularly, as well as additional fish oil supplements (15%), vitamin D (300–1000 IU) and/or vitamin E (>600 IU) (10%), or vitamin B12 (4%), for at least 1 year prior to brain imaging. This study examined nutrient intake from food sources only, showing significant associations between NPs and brain AD biomarkers in NL individuals. Other

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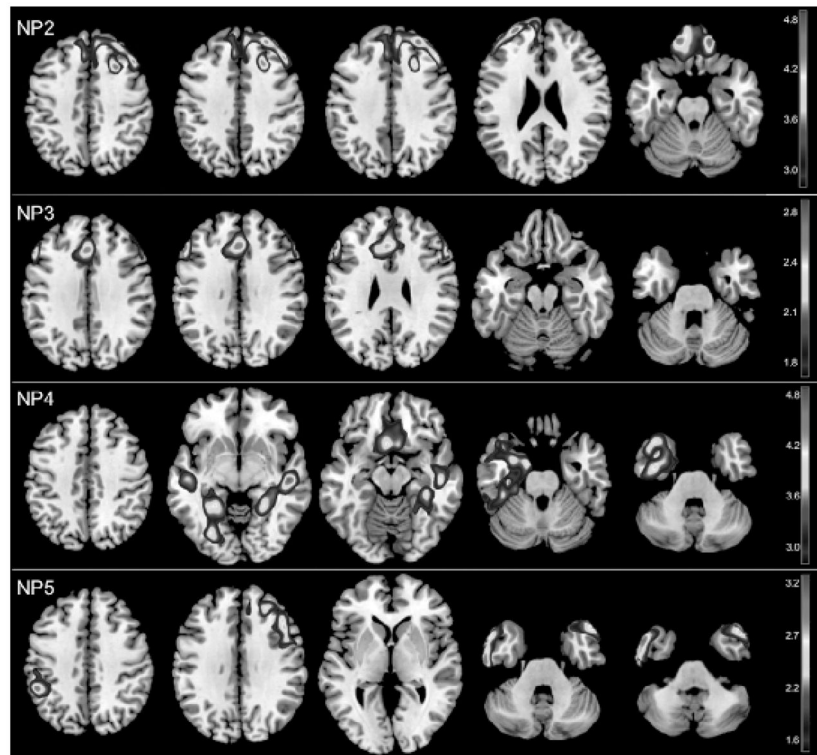


Figure 1. Statistical parametric maps (SPMs) showing associations between nutrient patterns (NPs) and brain glucose metabolism (METglc) on FDG-PET

Brain regions showing positive associations between METglc and (NP2) intake of vitamin E, monounsaturated and polyunsaturated fats (ω -3 and ω -6 PUFA); (NP3) intake vitamin A, vitamin C, carotenoids and dietary fibers; (NP4) intake of vitamin B12, vitamin D and zinc; brain regions showing negative associations between METglc and (NP5) intake of saturated, trans-saturated fats and sodium. SPMs are represented on a color-coded scale at $p < 0.001$, and displayed onto a standardized MRI. Results are adjusted for age, gender, education, BMI, APOE, family history and total caloric intake.

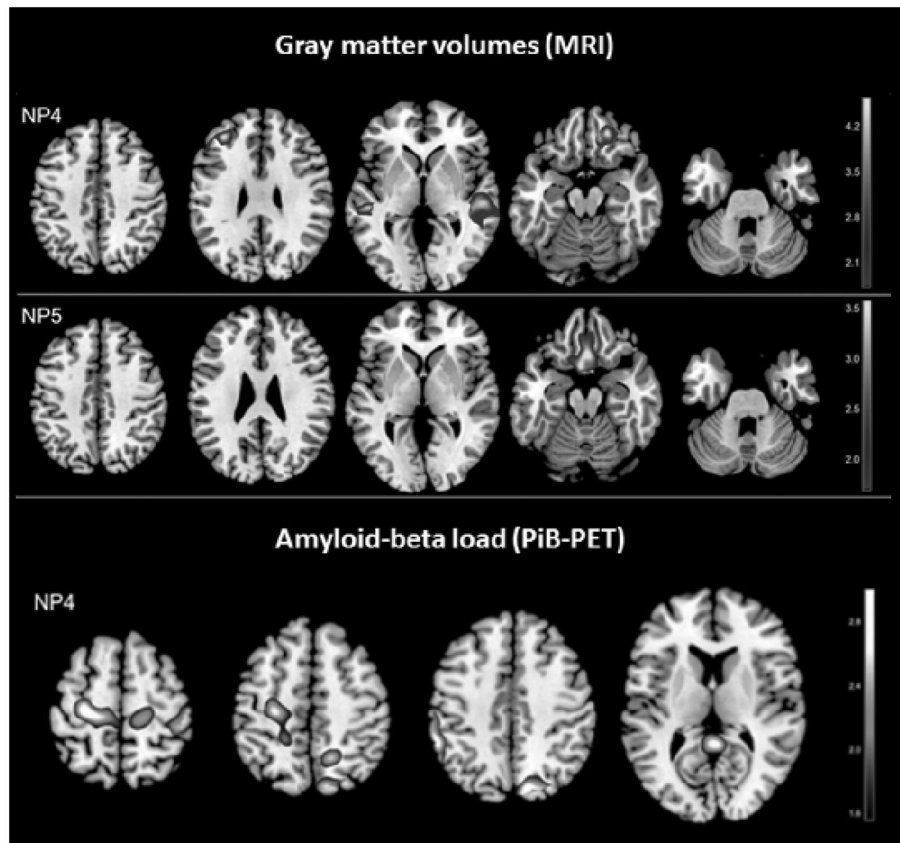


Figure 2. Statistical parametric maps (SPMs) showing significant regional associations between nutrient patterns (NPs), gray matter volumes (GMV) on MRI, and reduced brain amyloid load on PiB-PET

Top panel: brain regions showing positive associations between GMV and (NP4) intake of vitamin B12, vitamin D and zinc; brain regions showing negative associations between GMV and (NP5) intake of saturated, trans-saturated fats and sodium. Bottom panel: brain regions showing negative associations between PiB retention and (NP4) higher intake of vitamin B12, vitamin D and zinc. SPMs are represented on different color-coded scales at $p < 0.001$, and displayed onto a standardized MRI. Results are adjusted for age, gender, education, BMI, APOE, family history and total caloric intake.

Table 1

Demographic and clinical characteristics

N	52
Age, y, mean (SD)	54(11)
Gender, female/male (n, % female)	37/15 (71%)
Education, y, mean (SD)	16(2)
Family history of LOAD, % positive	79%
APOE ϵ 4 status, % positive	47%
Ethnicity (%)	
White	83%
Black	8%
Hispanic	6%
Other	3%
Body Mass Index [unitless], mean (SD)	25(4)
Hip to waist ratio [unitless], mean (SD)	1.3(0.6)
Blood pressure (mm/Hg), mean (SD)	
Systolic	119(14)
Diastolic	72(8)
Glucose (mg/dl)	78(12)
Cholesterol (mg/dl)	197(37)
HDL (mg/dl)	62(18)
LDL (mg/dl)	117(32)
Triglycerides (mg/dl)	89(39)
Homocysteine (micromol/l)	10(3)
Neuropsychological tests, mean (SD)	
Mini Mental State Exam	29(1)
Digit symbol substitution	63(10)
Paired associates delayed recall	7(2)
Paragraph delayed recall	10(3)
Designs	8(2)
Object naming	55(12)
Digit span, forward	7(13)
Digit span, backward	5(1)
WAIS-vocabulary	65(13)

Table 2

Nutrient pattern construction: Pattern structure and variance explained

Nutrients	Nutrient patterns (NP) ^a				
	NP1	NP2	NP3	NP4	NP5
Calcium	0.77 <i>b</i>		0.26		
Folic acid (B9)	0.79 <i>b</i>	0.33	0.44		
Iron	0.77 <i>b</i>	0.40	0.35		
Magnesium	0.67 <i>b</i>	0.54	0.33		
Niacin (B3)	0.81 <i>b</i>	0.39			
Phosphorus	0.80 <i>b</i>	0.47			
Potassium	0.60 <i>b</i>	0.46	0.49		
Pyridoxal phosphate (B6)	0.72 <i>b</i>	0.40	0.42		
Riboflavin (B2)	0.84 <i>b</i>				
Selenium	0.76 <i>b</i>	0.45			
Thiamin (B1)	0.88 <i>b</i>	0.31			
Beta Tocopherol	0.57	0.73 <i>b</i>			
Copper	0.38	0.63 <i>b</i>	0.39	0.34	-0.38
Delta Tocopherol	0.49	0.60 <i>b</i>			
Gamma Tocopherol	0.38	0.87 <i>b</i>			
Monounsaturated fats	0.38	0.82 <i>b</i>			
Omega3 PUFA		0.83 <i>b</i>		0.36	
Omega6 PUFA	0.33	0.92 <i>b</i>			
Polyunsaturated fats	0.32	0.92 <i>b</i>			
Vitamin E	0.36	0.84 <i>b</i>	0.28		
Alpha carotene			0.81 <i>b</i>		
Beta carotene			0.85 <i>b</i>		
Beta Cryptoxanthin			0.59 <i>b</i>		

Nutrients	Nutrient patterns (NP) ^a				
	NP1	NP2	NP3	NP4	NP5
Dietary Fibers	0.55	0.44	0.60 ^b		
Lutein and Zeaxanthin			0.80 ^b		
Vitamin A	0.32		0.79 ^b		
Vitamin C	0.45		0.74 ^b		
Calciferol (Vitamin D3)	0.35			0.86 ^b	
Cobalamin (B12)				0.95 ^b	
Lycopene	-0.23		0.33	0.40 ^b	0.30
Zinc		0.45		0.69 ^b	-0.32
Cholesterol			-0.24	0.31	0.75 ^b
Saturated fats	0.49				0.71 ^b
Sodium		0.55			0.72 ^b
Trans-saturated fats	0.43	0.31			0.56 ^b
% variance explained by each component	30	26	16	10	4
Cumulative % of variance explained with each extraction	30	56	72	82	86

^aLoading coefficients represent the actual correlations between each variable and the identified factor scores from PCA (i.e., patterns). Nutrient pattern interpretation is based on the strongest loading coefficients within each pattern. Each standardized summary score is a linear combination of the nutrients that mostly represent the respective pattern. Coefficients -0.20 and <0.2 were excluded to simplify the table and emphasize dominant nutrients within each pattern. As such, blank spaces reflect non-significant to weak associations;

^bConsidered the dominant nutrients in the pattern. For example, a high pattern NP1 score is interpreted as high intake of B-vitamins (B1, B2, B3, B6, B9) and most minerals (i.e., calcium, iron, magnesium, phosphorus, potassium and selenium). Nutrients within in each pattern are listed in alphabetical order.

Table 3
Brain regions showing significant relationships between NPs and brain glucose metabolism on FDG-PET

Cluster extent	x*	y	z	Z†	Anatomical region	Brodmann area
Positive associations between FDG uptake and NP2						
71	14	38	-19	3.61	Right Cerebrum, Frontal Lobe, Inferior Frontal Gyrus	11
54	-14	40	-19	3.51	Left Cerebrum, Frontal Lobe, Middle Frontal Gyrus	11
40	-25	48	30	3.48	Left Cerebrum, Frontal Lobe, Superior Frontal Gyrus	9
31	24	37	41	3.35	Right Cerebrum, Frontal Lobe, Middle Frontal Gyrus	8
39	24	14	45	3.32	Right Cerebrum, Frontal Lobe, Middle Frontal Gyrus	8
Positive associations between FDG uptake and NP3						
40	-54	27	28	2.71	Left Cerebrum, Frontal Lobe, Middle Frontal Gyrus	46
47	-2	26	31	2.57	Left Cerebrum, Limbic Lobe, Cingulate Gyrus	32
Positive associations between FDG uptake and NP4						
277	44	-21	-4	4.08	Right Cerebrum, Temporal Lobe, Superior Temporal Gyrus	22
259	-15	-3	-17	3.64	Left Cerebrum, Limbic Lobe, Parahippocampal Gyrus	34
278	-28	-1	-15	3.32	Left Cerebrum, Limbic Lobe, Parahippocampal Gyrus, Amygdala	
116	-36	14	-22	3.55	Left Cerebrum, Temporal Lobe, Superior Temporal Gyrus	38
141	-39	-5	-32	3.42	Left Cerebrum, Temporal Lobe, Middle Temporal Gyrus	21
59	-34	-8	-29	3.27	Left Cerebrum, Limbic Lobe, Uncus	20
68	-6	13	-10	3.41	Left Cerebrum, Frontal Lobe, Medial frontal Gyrus	25
333	23	-43	-5	3.39	Right Cerebrum, Limbic Lobe, Parahippocampal Gyrus	36
235	-29	-44	-8	3.32	Left Cerebrum, Limbic Lobe, Parahippocampal Gyrus	37
Negative associations between FDG uptake and NP5						
43	-47	8	-35	3.52	Left Cerebrum, Temporal Lobe, Middle Temporal Gyrus	21
44	-56	-10	-30	3.21	Left Cerebrum, Temporal Lobe, Inferior Temporal Gyrus	20
44	36	20	37	3.33	Right Cerebrum, Frontal Lobe, Middle frontal Gyrus	9
	-48	-44	39	2.87	Left Cerebrum, Parietal Lobe, Inferior Lobule	40
	42	13	-33	2.72	Right Cerebrum, Temporal Lobe, Superior Temporal Gyrus	38

* Coordinates (x, y, z) from Talairach and Tournoux.

† Z values at the peak of maximum significance at p<0.001, correcting for age and total caloric intake. Only contrasts yielding significant results are reported.

Abbreviations: NP1 = Vit B & minerals, NP2 = PUFA & Vit E, NP3 = Anti-oxidants & Fibers, NP4 = Vit B12 & D and zinc, NP5 = Fats and sodium

Table 4
Brain regions showing significant relationships between NPs and brain gray matter volumes (GMV) on MRI

Cluster extent	x*	y	z	Z †	Anatomical region	Brodman area
Positive associations between GMV and NP1						
62	-28	45	17	4.16	Left Cerebrum, Frontal Lobe, Superior Frontal Gyrus	10
Positive associations between GMV and NP4						
629	50	-20	-6	4.48	Right Cerebrum, Temporal Lobe, Superior Temporal Gyrus	22
	52	-25	6	3.95	Right Cerebrum, Temporal Lobe, Superior Temporal Gyrus	41
	44	-25	5	3.75	Right Cerebrum, Temporal Lobe, Superior Temporal Gyrus	22
51	-48	-25	1	4.17	Left Cerebrum, Temporal Lobe, Superior Temporal Gyrus	21
81	19	40	-17	4.15	Right Cerebrum, Frontal Lobe, Middle Frontal Gyrus	11
79	-34	32	32	4.09	Left Cerebrum, Frontal Lobe, Middle Frontal Gyrus	9
Negative associations between GMV and NP5						
123	-5	10	-16	3.64	Left Cerebrum, Frontal Lobe, Medial Frontal Gyrus	25
	0	22	-12	3.59	Left Cerebrum, Limbic Lobe, Anterior Cingulate, Gray Matter	32

* Coordinates (x, y, z) from Talairach and Tournoux.

† Z values at the peak of maximum significance at $p < 0.001$, correcting for age and total caloric intake. Only contrasts yielding significant results are reported.

Abbreviations: see legend to Table 3

Table 5
Brain regions showing significant relationships between NPs and brain A β load on PiB-PET

Cluster extent	x*	y	z	Z†	Anatomical region	Brodman area
Negative associations between PiB retention and NP4						
307	-14	-12	-42	2.90	Left Cerebrum, Limbic Lobe, Cingulate Gyrus	24
768	-29	25	50	2.79	Left Cerebrum, Frontal Lobe, Superior Frontal Gyrus	8
279	0	-43	8	2.78	Left Cerebrum, Limbic Lobe, Posterior Cingulate Gyrus	29
134	-23	-10	50	2.72	Left Cerebrum, Frontal Lobe, Superior Frontal Gyrus	8
30	7	-76	43	2.72	Right Cerebrum, Parietal Lobe, Precuneus	7
77	-53	-45	40	2.66	Left Cerebrum, Parietal Lobe, Inferior Parietal Lobule	40
85	12	-58	46	2.64	Right Cerebrum, Parietal Lobe, Precuneus	7
	37	-37	54	2.63	Right Cerebrum, Parietal Lobe, Inferior Parietal Lobule	40

* Coordinates (x, y, z) from Talairach and Tournoux.

† Z values at the peak of maximum significance at $p < 0.001$, correcting for age and total caloric intake. Only contrasts yielding significant results are reported.

Abbreviations: see legend to Table 3