Lifestyle and vascular risk effects on MRI-based biomarkers of Alzheimer’s disease: a cross-sectional study of middle-aged adults from the broader New York City area

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ABSTRACT

Objective To investigate the effects of lifestyle and vascular-related risk factors for Alzheimer’s disease (AD) on in vivo MRI-based brain atrophy in asymptomatic young to middle-aged adults.

Design Cross-sectional, observational.

Setting Broader New York City area. Two research centres affiliated with the Alzheimer’s disease Core Center at New York University School of Medicine.

Participants We studied 116 cognitively normal healthy research participants aged 30–60 years, who completed a three-dimensional T1-weighted volumetric MRI and had lifestyle (diet, physical activity and intellectual enrichment), vascular risk (overweight, hypertension, insulin resistance, elevated cholesterol and homocysteine) and cognition (memory, executive function, language) data. Estimates of cortical thickness for entorhinal (EC), posterior cingulate, orbitofrontal, inferior and middle temporal cortex were obtained by use of automated segmentation tools. We applied confirmatory factor analysis and structural equation modelling to evaluate the associations between lifestyle, vascular risk, brain and cognition.

Results Adherence to a Mediterranean-style diet (MeDi) and insulin sensitivity were both positively associated with MRI-based cortical thickness (diet: β≥0.26, insulin sensitivity β≥0.58, P≤0.008). After accounting for vascular risk, EC in turn explained variance in memory (P≤0.001). None of the other lifestyle and vascular risk variables were associated with brain thickness. In addition, the path associations between intellectual enrichment and better cognition were significant (β≥0.25, P≤0.001), as were those between overweight and lower cognition (β≥-0.22, P≤0.01).

Conclusions In cognitively normal middle-aged adults, MeDi and insulin sensitivity explained cortical thickness in key brain regions for AD, and EC thickness predicted memory performance in turn. Intellectual activity and overweight were associated with cognitive performance through different pathways. Our findings support further investigation of lifestyle and vascular risk factor modification against brain ageing and AD. More studies with larger samples are needed to replicate these research findings in more diverse, community-based settings.

INTRODUCTION

Alzheimer’s disease (AD) is the most common form of dementia, affecting nearly 34 million people worldwide.1 Unless effective strategies for prevention are found, the prevalence of AD is expected to triple by 2050.
There is substantial epidemiological evidence linking modifiable risk factors, such as midlife hypertension, obesity, diabetes, poor diet, physical and intellectual inactivity, with increased risk of late-onset AD. Recent population-attributable models estimated that one in every three cases of AD may be accounted for by these lifestyle and vascular risk factors in healthy middle-aged adults conducted at two research centres affiliated with the Alzheimer’s disease Coordinating Center of New York University (NYU) School of Medicine between 2010 and 2016. Details about the studies have previously been published.\(^6\) All participants provided informed consent to participate in this study.

Given the current lack of disease-modifying treatment,\(^2\) and increasing awareness that AD pathology develops over many years prior to clinical symptoms,\(^3\) the potential for prevention is crucial to reduce AD risk and/or delay the onset of cognitive decline.

There are substantial epidemiological evidence linking modifiable risk factors, such as midlife hypertension, obesity, diabetes, poor diet, physical and intellectual inactivity, with increased risk of late-onset AD. Recent population-attributable models estimated that one in every three cases of AD may be accounted for by these lifestyle and vascular-related risk factors.\(^1\)\(^14\) As a result, trends in AD risk reduction research have focused on lifestyle interventions as well as vascular risk reduction.\(^5\)\(^2\)\(^3\)\(^3\)

### METHODS

#### Participants

Participants were enrolled in observational brain imaging studies of clinically and cognitively normal middle-aged adults conducted at two research centres affiliated with the Alzheimer’s disease Coordinating Center of New York University (NYU) School of Medicine between 2010 and 2016. Details about the studies have previously been published.\(^6\)\(^7\)\(^17\)\(^-\)\(^21\) All participants provided informed consent to participate in this study.

### Table 1 Participants’ demographic and clinical characteristics

| Characteristic                  | n  | 116 |
|---------------------------------|----|-----|
| Age, years                      | 50 (6), range 30–60 | |
| Sex, % female                   | 62 |     |
| Education, years                | 16 (2), range 12–22 | |
| Family history of LOAD, % positive | 38 |     |
| APOE ε4 carriers, % positive    | 40 |     |
| Ethnicity, % white              | 68 |     |
| Subjective complaints, % positive | 44 |     |
| Laboratory findings             |    |     |
| BMI (kg/cm\(^2\))               | 25 (4), range 18–37 | |
| Hypertension, % positive        | 14 |     |
| QUICKI score (unitless)         | 0.32 (0.03) | |
| Plasma cholesterol/HDL ratio (unitless) | 3.3 (0.8) | |
| Plasma homocysteine (μmol/L)    | 7.9 (6.2) | |
| Lifestyle measures              |    |     |
| Mediterranean diet (unitless)   | 4.3 (1.9), range 1–8 | |
| Physical activity (metabolic equivalent/hour) | 9.8 (6.3), range 1–37 | |
| Intellectual activity (unitless) | 3.7 (0.7), range 1.8–4.9* | |

Values are mean (SD) unless otherwise specified. *Data were not available for 20 out of 116 patients.

APOE, apolipoprotein E; BMI, body mass index; HDL, high-density lipoprotein; LOAD, late-onset Alzheimer’s disease; QUICKI, Quantitative Insulin Sensitivity Check Index.

Neuropathologically, AD is characterised by the presence of amyloid-beta (Aβ) plaques, neurofibrillary tangles and neuronal loss in selectively vulnerable brain regions.\(^6\) Neuronal loss in AD originates in the medial temporal lobes during the normal stages of cognition and spreads to cortical regions with disease progression.\(^15\) MRI has long been used to visualise neurodegeneration in vivo. Several MRI studies have shown that brain cortical thinning (ie, atrophy) in AD-vulnerable regions begins many years prior to dementia onset, making this biomarker an ideal candidate to monitor the effectiveness of preventative interventions.\(^16\)

This study examines the simultaneous effects of multiple modifiable lifestyle and vascular risk factors on MRI-based cortical atrophy in a cohort of middle-aged healthy adults at risk for AD.
E (APOE) genotypes were determined using standard quantitative PCR procedures.17

Cognitive measures
The neuropsychological battery of tests was previously described.22 Briefly, three cognitive domains were assessed from the following tests: memory (immediate and delayed recall of a paragraph and immediate and delayed recall of paired associates), executive function (Wechsler Adult Intelligence Scale (WAIS) digit symbol substitution) and language (WAIS vocabulary).

We modelled memory as a latent factor composed of standardised versions of immediate and delayed recall scores using confirmatory factor analysis (CFA) (alpha=0.84). The three cognitive domains were examined as outcome variables (see Statistical analysis).

Vascular risk measures
Vascular risk factors included in the model were: (1) overweight, measured using the body mass index (BMI, kg/m²); (2) presence of hypertension, conservatively determined based on either current antihypertensive treatment or blood pressure assessments; (3) elevated plasma cholesterol and/or elevated plasma homocysteine, as obtained after overnight fasting using standard laboratory procedures and (4) insulin sensitivity, measured with the Quantitative Insulin Sensitivity Check Index,23 where lower scores reflect greater insulin resistance and higher scores reflect insulin sensitivity.

Lifestyle measures
Diet
Dietary data regarding average food consumption over the prior year were obtained using the Block or Harvard food frequency questionnaire, as previously described.19 Briefly, food items were categorised into food groups based on similarities in food and nutrient composition, and intake of each food group was calculated by summing the intakes of food group items. Mediterranean diet (MeDi) scores were generated as the sum of caloric intake-adjusted daily gram intake dichotomised based on either current antihypertensive treatment or blood pressure assessments; (3) elevated plasma cholesterol and/or elevated plasma homocysteine, as obtained after overnight fasting using standard laboratory procedures and (4) insulin sensitivity, measured with the Quantitative Insulin Sensitivity Check Index,23 where lower scores reflect greater insulin resistance and higher scores reflect insulin sensitivity.

Lifestyle measures
Diet
Dietary data regarding average food consumption over the prior year were obtained using the Block or Harvard food frequency questionnaire, as previously described.19 Briefly, food items were categorised into food groups based on similarities in food and nutrient composition, and intake of each food group was calculated by summing the intakes of food group items. Mediterranean diet (MeDi) scores were generated as the sum of caloric intake-adjusted daily gram intake dichotomised.
relative to the sex-specific median for each beneficial (fruits, vegetables, legumes, cereals and fish; monounsaturated to saturated fat ratio; mild to moderate alcohol consumption) versus detrimental (meat and dairy products) component. Greater score indicate greater MeDi adherence.

Physical activity

The Baecke and Minnesota leisure time physical activity questionnaires were used to estimate the current level of physical activity. For each activity, information was collected on the frequency and duration of engagement over the past year. Frequency and duration data were multiplied with an activity-specific intensity code indicating calorie expenditure. We standardised and summed the activity-dependent scores from each test to obtain the overall annual intensity of physical activity per person, which was converted to metabolic equivalents.

Table 2  Results of structural equation modelling for model 1: lifestyle, vascular risk and brain

| Associations between lifestyle and brain MRI biomarkers | β, SE | P value |
|--------------------------------------------------------|-------|---------|
| Diet → Brain                                           | 0.260 | 0.099  | 0.009 |
| Diet → Brain                                           | 0.204 | 0.086  | 0.018 |
| Physical activity → Brain                              | −0.130| 0.108  |       |
| Physical activity → Brain                              | −0.058| 0.047  |       |
| Intellectual enrichment → Brain                         | −0.192| 0.098  |       |
| Intellectual enrichment → Brain                         | −0.190| 0.102  |       |

| Associations between vascular risk and brain MRI biomarkers | β, SE | P value |
|---------------------------------------------------------------|-------|---------|
| Overweight → Brain                                             | −0.067| 0.101  |       |
| Overweight → Brain                                             | −0.116| 0.173  |       |
| Insulin sensitivity → Brain                                    | 0.559 | 0.081  | <0.001|
| Insulin sensitivity → Brain                                    | 0.029 | 0.006  | <0.001|
| Plasma cholesterol/HDL → Brain                                 | −0.038| 0.101  |       |
| Plasma homocysteine → Brain                                    | 0.177 | 0.315  |       |
| Plasma homocysteine → Brain                                    | 0.175 | 0.372  |       |
| Hypertension → Brain                                            | 0.076 | 0.106  |       |
| Hypertension → Brain                                            | 0.032 | 0.042  |       |

The table consists of standardised betas (β), their SEs and P values of the estimates from the full model with age entered as a covariate. The βs can be interpreted as partial correlations. Only significant P values (two tailed) are reported. Paths in italics are adjusted for sex and APOE status as covariates.

Intellectual enrichment

Intellectual activity and years of education were used as measures of intellectual enrichment as described below. Intellectual activity throughout life was assessed using a validated 25-item interview in which participants were asked to report how often they engaged in common cognitively demanding activities that depend minimally on socioeconomic status, such as reading books or newspapers, writing letters or e-mails, going to the library and playing games at different age epochs. Previous studies described this instrument in detail and reported high internal consistency and positive associations of intellectual activity with educational and cognitive performance.

MRI acquisition and processing

All subjects received three-dimensional volumetric T1-weighted MRI on a 3T scanner according to published protocols. MRIs were acquired and preprocessed as described. Volumetric segmentation, cortical surface reconstruction and parcellation were performed using the FreeSurfer V.5.3 software package. Cortical thickness measures were obtained for a subset of a priori-defined regions of interest (ROIs) known to show atrophic changes early in AD and in association with lifestyle among the four sets of variables: lifestyle (diet, physical activity and intellectual enrichment), vascular-related risk (overweight, hypertension, insulin resistance, elevated plasma cholesterol and homocysteine levels), MRI biomarkers (latent variable of brain structure) and cognitive measures (latent variable of memory, executive function and language).

Statistical analysis

We constructed structural equation models (SEMs) (MPlus V.7) to evaluate interdependent relationships among the four sets of variables: lifestyle (diet, physical activity and intellectual enrichment), vascular-related risk (overweight, hypertension, insulin resistance, elevated plasma cholesterol and homocysteine levels), MRI biomarkers (latent variable of brain structure) and cognitive measures (latent variable of memory, executive function and language). The three cognitive domains were modelled as a function of lifestyle variables, vascular variables and MRI biomarkers (see online supplementary efigure 1). Lifestyle and vascular variables were specified as exogenous variables. Brain structure formed an intermediate layer of endogenous variables, and the cognitive domains were the final downstream endogenous variables. As such, exogenous variables (lifestyle and vascular risk measures) predicted brain variables, and brain variables in turn predicted cognition. We regressed brain structure on the lifestyle and vascular variables. We then regressed...
cognitive domains on all other variables in the model. Additionally, ran a separate model in which we replace the latent variable of brain structure with the two limbic regions (ie, EC and PCC) based on evidence of their earlier predictive capacity.

Since variables in the two exogenous domains, lifestyle and vascular risk, were only marginally related to each other (see Results), the model was further broken down into the following three submodels: (1) we first modelled the relationships between lifestyle, vascular risk and brain, excluding cognition from the model; (2) we modelled the relationships between lifestyle, brain and cognition, excluding vascular risk and (3) we modelled the relationships between vascular risk, brain and cognition, excluding lifestyle from the model.

All models were adjusted for the intercorrelations between lifestyle variables, vascular-risk variables, brain structure and cognitive measures as appropriate. Age, gender and APOE status were examined as covariates. Brain measures were evaluated with and without adjustment for TIV.

For each analysis, we first fit the full models, estimating all paths and then assessed the reduced models suggested by the primary analysis (eg, without significant paths removed). We used $\chi^2$ statistic, comparative fit index (CFI) and root mean square error of approximation (RMSEA) goodness of fit tests to indicate model fits for the full models. Each of these measures incorporates unique criteria to assess fit so a summary of all three measures provides a more comprehensive estimate. After the path coefficients were derived, the paths were thresholded to achieve a second, more parsimonious model, by eliminating paths with P values >0.05. Path elimination was monitored via the $\chi^2$, CFI and RMSEA. Good model fit can be reflected by a $\chi^2$ to df ratio $\leq 2.0$, a CFI $\geq 0.85$ and an RMSEA $<0.06$. The CFI and RMSEA are among the measures least affected by sample size and perform very well at all sample sizes. As such, CFI and RMSEA were our primary reporting criteria for the reduced models.

Some of the reduced models were saturated (eg, these models accounted for all possible relationships that could exist among variables in our dataset), making overall model fit statistics not applicable. In this case, we focused on specific paths and path significance rather than omnibus measures model fits. The standardised beta coefficients ($\beta$) can be interpreted as partial correlations. Given that all of the paths are standardised, one can judge meaningfulness by their raw and comparative path weights. All analyses used maximum likelihood estimation in the MPlus package.

RESULTS

Participants

A total of 116 participants were included in the analysis (table 1). Participants were on average 50 years old (range 25–60), 62% women, with education $\geq 12$ years. Of these, 38% had a family history of AD, and 40% had at least one copy of the APOE4 allele (APOE4 carriers).

Structural equation modelling

In the full model, age was associated with risk of hypertension ($\beta=0.18$, P=0.049) and, although it tended to be negatively related to brain structure ($\beta=−0.10$ to −0.19, ns), it was not significantly associated with the lifestyle or cognitive variables.

Men, as expected, showed higher vascular risk than women, as reflected in a higher frequency of hypertension, elevated cholesterol and homocysteine levels (P=0.04). Women showed better executive function performance than men (P=0.001). Sex was not significantly or marginally related to lifestyle variables or brain structure.

APOE status was not associated with lifestyle or cognitive variables but was associated with plasma cholesterol (P=0.033) and marginally associated with brain structure (P=0.07). These effects were driven by APOE4 carriers exhibiting higher cholesterol levels and reduced cortical thickness than non-carriers.

Given these relationships, all subsequent analyses were performed with and without adjusting for age, sex and APOE, as appropriate.

Among lifestyle variables, with and without controlling for the above confounds, participants with higher intellectual enrichment showed higher MeDi adherence ($\beta=0.27$, P≤0.028). Otherwise, lifestyle variables were not significantly related to one another.

Among vascular variables, BMI was negatively associated with insulin sensitivity ($\beta=−0.27$, P=0.01) and positively associated with cholesterol levels and hypertension ($\beta=0.32$ and $\beta=0.29$, P≤0.001). Cholesterol levels and hypertension were associated with each other ($\beta=0.24$, P=0.05).

Vascular and lifestyle variables were not associated with each other, except for higher intellectual enrichment correlating with lower risk of hypertension ($\beta=−0.22$, P≤0.05), and insulin sensitivity was marginally, though not significantly, associated with diet ($\beta=0.12$, P=0.18). The model was therefore broken down in the following three submodels:

Lifestyle, vascular risk and MRI-based biomarkers

Figure 1 represents the path diagrams from model 1, which exhibited adequate fit, $\chi^2(41)=95.37$, RMSEA=0.05, CFI=0.85, P≤0.001. Controlling for age, diet and insulin sensitivity were the only factors positively associated with brain structure ($\beta=0.26$ and $\beta=0.58$, respectively, P≤0.008, table 2). The reduced model fit the data well (RMSEA <0.01, CFI=0.90) reflecting the fact that diet and insulin sensitivity both significantly and independently predicted brain structure (diet: $\beta=0.20$, P=0.017 and insulin sensitivity: $\beta=0.57$, P=0.001), consistent with the pattern of significant and non-significant paths obtained in full model 1. Including sex and APOE status as covariates did not significantly influence these relationships (table 2).
Restricting analysis to limbic structures fit the data well (RMSEA <0.05, CFI=1.0), confirming that insulin sensitivity was positively associated with both EC and PCC ($\beta_s=0.42$ and $\beta_s=0.38$, $P<0.001$), and diet was associated with PCC ($\beta_s=0.23$, $P=0.004$) and marginally with EC ($\beta_s=0.13$, $P=0.12$) (figure 2).

**Lifestyle, MRI-based biomarkers and cognition**

Model 2 exhibited adequate fit, $\chi^2_{(64)}$=108.51, RMSEA <0.05, CFI=0.90, $P<0.001$. The reduced models were completely saturated (eg, all possible relationships that could exist among variables in the dataset were accounted for). As such, we focused on specific paths and path significance rather than omnibus measures of model fit.

As in model 1, with and without adjusting for age and TIV, diet was the only lifestyle factor positively associated with brain structure ($\beta_s=0.27$, $P=0.002$, **table 3** and **figure 1**). Diet and physical activity were not associated with cognitive function, whereas intellectual enrichment was positively associated with cognition ($P<0.05$, **table 3**). Specifically, intellectual activity was positively associated with memory and executive function ($\beta_s=0.28$ and $\beta_s=0.39$, respectively, $P<0.02$), and marginally with language ($\beta_s=0.25$, $P=0.08$). Education positively predicted memory, executive function and language ($\beta_s=0.21$–0.33, $P=0.038$).

Brain structure was not significantly associated with cognition. However, when analyses were restricted to limbic structures, diet showed positive associations with PCC ($\beta_s=0.27$, $P=0.006$) and EC ($\beta_s=0.17$, $P=0.08$), which in turn positively, though marginally, predicted memory ($\beta_s=0.19$, $P=0.078$) (**figure 2**).

Including sex and APOE status as covariates did not influence the other relationships of interest in any of the models (**table 3**).

**Vascular risk, MRI-based biomarkers and cognition**

Model three exhibited adequate fit, $\chi^2_{(78)}$=151.54, RMSEA <0.05, CFI=0.89, $P<0.001$. The reduced models were completely saturated and overall model fit statistics were not applicable. Below, we focus on path significance rather than omnibus measures of model fit.

As in model 1, insulin sensitivity was the only vascular factor associated with brain structure ($\beta_s=0.58$, $P<0.01$, **table 4** and **figure 1**). Brain structure was not significantly associated with cognition. However, when analysis was restricted to limbic structures, insulin sensitivity showed positive associations with EC ($\beta_s=0.30$, $P<0.001$) and EC in turn positively predicted memory ($\beta_s=0.39$, $P<0.001$) (**figure 2**). Additionally, BMI was negatively associated with memory and executive function ($\beta_s=0.22$ and $\beta_s=0.27$, $P=0.021$, **figure 2**).

Including sex and APOE status as covariates did not significantly influence the other relationships of interest (**table 4**).
Table 3  Results of structural equation modelling for model 2: lifestyle, brain and cognition

| Path                        | β  | SE   | P value |
|-----------------------------|----|------|---------|
| Diet → Brain                | 0.253 | 0.098 | 0.010   |
| Diet → Brain                | 0.253 | 0.099 | 0.010   |
| Physical activity → Brain   | −0.129 | 0.107 |         |
| Physical activity → Brain   | −0.132 | 0.126 |         |
| Intellectual enrichment → Brain | −0.113 | 0.125 |         |
| Intellectual enrichment → Brain | −0.164 | 0.099 |         |
| Intellectual enrichment → Brain | −0.168 | 0.099 |         |
| Education → Brain           | −0.112 | 0.124 |         |
| Education → Brain           | −0.168 | 0.099 |         |

Table 4  Results of structural equation modelling for model 3: vascular risk, brain and cognition

| Path                        | β  | SE   | P value |
|-----------------------------|----|------|---------|
| Overweight → Brain          | −0.067 | 0.101 |         |
| Overweight → Brain          | −0.116 | 0.173 |         |
| Insulin sensitivity → Brain | 0.559 | 0.081 |         |
| Insulin sensitivity → Brain | 0.029 | 0.006 | <0.001  |
| Plasma cholesterol/HDL → Brain | −0.038 | 0.101 |         |
| Plasma cholesterol/HDL → Brain | −0.010 | 0.037 |         |
| Plasma homocysteine → Brain | 0.177 | 0.315 |         |
| Plasma homocysteine → Brain | 0.175 | 0.372 |         |
| Hypertension → Brain        | 0.076 | 0.106 |         |
| Hypertension → Brain        | 0.032 | 0.042 |         |

The table consists of standardised betas (βs), their SEs, and P values of the estimates from the full model with age entered as a covariate. The βs can be interpreted as partial correlations. Only significant P values (two tailed) are reported. Paths in italics are adjusted for sex and APOE status as covariates.
Discussion

The major conclusions of this study were the following: (1) among lifestyle and vascular risk factors, diet and insulin sensitivity explained variability in brain cortical thickness in cognitively healthy, middle-aged adults; (2) MRI measures of limbic structures in turn explained variability in memory performance; (3) intellectual enrichment and increased BMI explained variability in cognitive performance through different pathways.

Our results indicate that diet and insulin sensitivity may be among the earliest modifiable risk factors to influence the expression of AD biomarkers, suggesting these modifiable risk factors may alter risk of AD pathophysiology during middle age. These associations were independent of age, sex and APOE genotype.

Findings of associations between lower MeDi adherence and increased brain atrophy are consistent with the literature in the elderly. Likewise, we and others have described strong associations between insulin resistance and limbic cortex atrophy in both adolescents and non-demented elderly. Our data extend prior observations to a population of middle-aged healthy adults and offers a comprehensive view of how the interplay of lifestyle and vascular factors influences possible AD risk. Indeed, besides being linked with lower risk of dementia, the MeDi was shown to support healthier insulin regulation and cardiovascular health. Given the known increased AD risk associated with prediabetes and type 2 diabetes, our findings fit with increased cardiovascular risk being a driver of increased brain ageing and AD.

The statistical approach we employed allowed us to simultaneously examine several lifestyle and vascular risk factors and discern their independent as well as intercorrelated effects on brain atrophy and cognition. Other studies that used SEM reported minimal or null associations between physical and intellectual activities and AD biomarkers in non-demented elderly, although biomarkers independently predicted cognitive function. These studies, however, did not take into account diet or vascular risk. Our study in a younger cohort (mean age 50 vs 78–82 years) indicates that physical and intellectual activity do not impact brain ageing as much as diet does during the normal stages of cognition. When vascular risk factors were accounted for in the model, insulin sensitivity influences cortical thickness and its effects on limbic regions had significant effects on memory performance. These effects were present accounting for the impact of exercise, intellectual activity and additional vascular risk factors.

In our cohort, insulin resistance was strongly associated with increased BMI (ie, overweight). Although BMI was not directly associated with MRI measures, it negatively impacted memory and executive function through other pathways. It has long been known that midlife obesity affects cognitive performance, possibly by promoting cardiovascular disease (CVD) and Aβ deposition.

Additionally, intellectual enrichment was associated with better cognitive performance, suggesting a protective effect on AD risk. It is possible that continued intellectual stimulation may lower the risk or delay the onset of AD by enhancing cognitive reserve, as suggested by studies in the elderly.

Several issues require further consideration. First, our results are cross-sectional and do not allow for determination of causality or temporal relationships between lifestyle, vascular risk, brain biomarkers and cognitive status. Studies with larger samples and longitudinal follow-ups are needed to examine lifestyle and vascular risk factors as possible AD risk modifiers. From a statistical perspective, even though our model fits were adequate and the observed relationships were supported by the data, the associations were not so strong to yield an ideal model fit. To better evaluate the impact of modifiable risk factors on brain ageing, we performed additional linear regression analyses to estimate R-squared values for the predictors identified in the reduced SEM models. R-squared is the ‘per cent of variance explained’ by the model (eg, the fraction by which the variance of the errors is less than the variance of the dependent variable). As with SEM analysis, diet and insulin sensitivity were both significant predictors of brain structure, yielding a combined R-squared=0.28 (age adjusted), P<0.001. In other words, the combination of diet and insulin sensitivity explained 28% of the variance in brain MRI measures. Additionally, intellectual enrichment and BMI explained 11% and 8% of the variance in cognitive measures, respectively (P≤0.022). As such, these modifiable risk factors were fairly good predictors after adjusting for all other variables in the model. The associations were nonetheless modest, which is not unexpected in a relatively young, healthy population. That said, caution should be exerted in interpreting these data. For instance, some of the null associations may depend on sample size limitations. However, null effects observed in our cohort are consistent with negative findings from large-scale studies in the elderly. Therefore, we offer that the strongest arguments of the study are the significant findings which manifest themselves despite these limitations. Overall,
our data suggest that lifestyle and vascular risk have an impact on brain ageing during midlife, thus supporting further investigation of modifiable risk factors for healthy brain ageing and AD risk reduction.

As with other studies in asymptomatic at-risk individuals, imaging biomarkers were only modestly associated with cognitive measures, most likely because our patients were all cognitively normal and younger than 60. Previous studies have demonstrated that associations between brain biomarkers and cognition are evident in clinical populations, such as those with clear brain pathology but not among normal populations. Nonetheless, associations between limbic MRI measures and memory were significant, indicating that modifiable risk factors may impact cognitive health before old age.

We cannot exclude that our screening criteria may have biased effect estimates. For instance, participation was limited to healthy, middle-aged individuals without severe cardiac and CVD. Our goal was to identify possible vascular, metabolic and lifestyle correlates of brain health, prior to severe disease and at a young enough age for potential interventions to be impactful. However, this makes our population restricted in the variability of cardiometabolic disease. As such, our results are only relevant to middle-aged, cognitively normal, healthy men and women without active CVD, stroke or diabetes. Nonetheless, our results are in keeping with prospective, community-based studies of non-diabetic populations showing that insulin resistance and increased blood sugar levels, even at levels considered normal in standard glucose tests, increase AD risk.

Another limitation of the study pertains to the characterisation of lifestyle habits. We used self-report questionnaires of diet and lifestyle, which are vulnerable to error and may not have captured relevant dimensions of lifestyle activities that influence cognitive functioning. Additionally, given the cross-sectional nature of our study and synchronous timing of lifestyle and MRI assessments, we cannot exclude the possibility that dietary adherence or physical activity levels were short-term choices in this cohort, which are less likely to impact brain biomarkers than long-term choices. However, 90% of participants reported living the lifestyle assessed in the surveys for 5 years or more.

Although MRI measures of neurodegeneration are sensitive to early AD changes, they are believed to emerge after changes in neuronal activity and downstream to Aβ accumulation. As such, MRI measures are not specific to AD and offer limited information on whether the observed associations were due to AD or to other causes of cortical thinning. Amyloid and tau biomarkers specific to AD are warranted to investigate the potential of lifestyle and vascular risk intervention for AD prevention. However, Aβ deposition is an age-dependent phenomenon, with 0% of cognitively normal individuals aged 45–49 years and less than 6% of those in the fifth decade of life testing positive for Aβ. Considering that all our participants were cognitively normal and between 30 and 60 years of age, very few (if any) would have had substantial amyloid burden, making this cohort an ideal population for testing of primary prevention strategies.

We caution that present results were found in small numbers of carefully screened patients who were evaluated under controlled clinical conditions. Longitudinal studies with larger samples are needed to replicate and assess the generalisability of these preliminary findings in community-based populations with higher variability in socioeconomic and medical status, as well as to incorporate other AD biomarkers. Should preventative studies prove successful, work will be needed to estimate the effects of increased longevity on dementia burden in such an increasingly older, although healthier, population.

In conclusion, our results suggest that a well-rounded lifestyle that incorporates a healthy diet (such as the MeDi), reduces vascular risk factors (especially insulin resistance and overweight) and promotes intellectual activity might be neuroprotective during ageing. More studies are needed to evaluate midlife lifestyle and vascular risk factor modification for AD prevention.

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Funding This study was supported by NIH/NIA grants AG035137, AG13616, 2P01AG026572 and NIH/NIDDK DK083537; funding from the Department of Neurology at Weill Cornell Medical College and philanthropic support of the Alzheimer’s Prevention Clinic, Weill Cornell Medicine Disorders Program.

Competing interests None declared.

Patient consent Obtained.

Ethics approval This study was approved by NYU School of Medicine Institutional Review Board.

Provenance and peer review Not commissioned; externally peer reviewed.

Data sharing statement All relevant data have been included in the paper. Technical appendix, statistical code and dataset will be made available on request.

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