

Genetic Variation and Neuroimaging Measures in Alzheimer Disease

Alessandro Biffi, MD; Christopher D. Anderson, MD; Rahul S. Desikan, MD, PhD; Mert Sabuncu, PhD; Lynelle Cortellini, MSc; Nick Schmansky, MSc; David Salat, PhD; Jonathan Rosand, MD, MSc; for the Alzheimer's Disease Neuroimaging Initiative (ADNI)

Objective: To investigate whether genome-wide association study (GWAS)-validated and GWAS-promising candidate loci influence magnetic resonance imaging measures and clinical Alzheimer's disease (AD) status.

Design: Multicenter case-control study of genetic and neuroimaging data from the Alzheimer's Disease Neuroimaging Initiative.

Setting: Multicenter GWAS.

Patients: A total of 168 individuals with probable AD, 357 with mild cognitive impairment, and 215 cognitively normal control individuals recruited from more than 50 Alzheimer's Disease Neuroimaging Initiative centers in the United States and Canada. All study participants had *APOE* and genome-wide genetic data available.

Main Outcome Measures: We investigated the influence of GWAS-validated and GWAS-promising novel AD loci on hippocampal volume, amygdala volume, white

matter lesion volume, entorhinal cortex thickness, parahippocampal gyrus thickness, and temporal pole cortex thickness.

Results: Markers at the *APOE* locus were associated with all phenotypes except white matter lesion volume (all false discovery rate-corrected P values $< .001$). Novel and established AD loci identified by prior GWASs showed a significant cumulative score-based effect (false discovery rate $P = .04$) on all analyzed neuroimaging measures. The GWAS-validated variants at the *CRI* and *PICALM* loci and markers at 2 novel loci (*BIN1* and *CNTN5*) showed association with multiple magnetic resonance imaging characteristics (false discovery rate $P < .05$).

Conclusions: Loci associated with AD also influence neuroimaging correlates of this disease. Furthermore, neuroimaging analysis identified 2 additional loci of high interest for further study.

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Author Affiliations: Center for Human Genetic Research, Department of Neurology, Massachusetts General Hospital, Boston (Drs Biffi, Anderson, and Rosand and Ms Cortellini); Athinoula A. Martinos Center for Biomedical Imaging, Department of Radiology, Massachusetts General Hospital, Charlestown (Drs Desikan, Sabuncu, and Salat, Ms Cortellini, and Mr Schmansky); and Program in Medical and Population Genetics, Broad Institute, Cambridge, Massachusetts (Drs Biffi, Anderson, and Rosand and Ms Cortellini).

Group Information: A list of the ADNI investigators appears at http://www.loni.ucla.edu/ADNI/Collaboration/ADNI_Manuscript_Citations.pdf.

LATE-ONSET ALZHEIMER DISEASE (AD) is the most common cause of dementia and the fifth leading cause of death in Americans older than 65 years.¹ The mechanisms underlying AD onset and progression remain largely unexplained. A study of twins² has demonstrated a significant role for genetics in late-onset AD, with heritability estimates of 60% to 80%. Until recently, the only genetic variant consistently shown to influence AD risk and age at onset was *APOE* (OMIM 107741).³ New findings from genome-wide association studies (GWASs) identified 3 additional loci conferring risk for AD: *CLU* (OMIM 185430), *PICALM* (OMIM 603025), and *CRI* (OMIM 120620).^{4,5} Other promising loci were also reported in these GWASs but did not achieve P values sufficient for genome-wide significance.

Multiple neuroimaging measures correlate with AD risk and progression. These measures also appear to have genetic underpinnings, with heritability estimates ranging from 40% to 80%,⁶ and have been pro-

posed as surrogate end points in biological research and clinical trials in AD.^{7,8} The demonstration that recently discovered genetic risk factors for AD also influence these neuroimaging traits would provide important confirmation of a role for these genetic variants and suggest mechanisms through which they might be acting.

For editorial comment see page 663

We therefore investigated the genetics of AD-related neuroimaging measures using data collected as part of the Alzheimer's Disease Neuroimaging Initiative (ADNI). We investigated whether GWAS-validated and GWAS-promising candidate loci influence magnetic resonance imaging (MRI) measures and clinical status (cognitively normal, mild cognitive impairment [MCI] without progression to probable AD, MCI with progression to probable AD, and probable AD). Because of limitations in sample size and hence study power, we performed individual single-nucleotide polymorphism (SNP)-

based analyses and cumulative score-based analysis, which incorporated information from a collection of candidate SNPs.

METHODS

ALZHEIMER'S DISEASE NEUROIMAGING INITIATIVE

Participants were selected from the ADNI database (<http://www.loni.ucla.edu/ADNI>). The ADNI is a large, multisite, collaborative effort launched in 2003 by the National Institute on Aging, the National Institute of Biomedical Imaging and Bioengineering, the US Food and Drug Administration, private pharmaceutical companies, and nonprofit organizations as a public-private partnership aimed at testing whether serial MRI, positron emission tomography, other biological markers, and clinical and neuropsychological assessment can be combined to measure the progression of MCI and early AD. The principal investigator of ADNI is Michael Weiner, MD. ADNI is the product of many coinvestigators from a broad range of academic institutions and private corporations, with patients recruited from more than 50 sites across the United States and Canada. For more information, see <http://www.adni-info.org>. Data from the ADNI cohort were not used in either of the prior AD GWASs.^{4,5}

STUDY PARTICIPANTS

Participants were screened, enrolled, and followed up prospectively according to the ADNI study protocol described in detail elsewhere.⁹ The degree of clinical severity for each participant was evaluated by an annual semistructured interview. This interview generated an overall Clinical Dementia Rating (CDR) score and the CDR Sum of Boxes.¹⁰ The Mini-Mental State Examination¹¹ and a neuropsychological battery were also conducted.

Participants were selected from the ADNI database if they were classified at baseline as (1) cognitively normal control individuals with a CDR score of 0; (2) patients with MCI with Mini-Mental State Examination scores between 24 and 30, a subjective memory complaint verified by an informant, objective memory loss as measured by education-adjusted performance on the Logical Memory II subscale (delayed paragraph recall) of the Wechsler Memory Scale-Revised,¹² a CDR score of 0.5, absence of significant levels of impairment in other cognitive domains, essentially preserved activities of daily living, an absence of dementia at the time of the baseline MRI scan, and classified as having the amnesic subtype of MCI based on the revised MCI criteria¹³; and (3) patients with AD who met criteria for probable AD¹⁴ (CDR score of 1).

Among 746 study participants who fulfilled quality control criteria for genotype data, 171 qualified for an AD diagnosis at baseline, 364 had MCI, and 205 were cognitively normal controls. Among 364 with baseline MCI, longitudinal follow-up identified 140 who converted to an AD diagnosis and 217 who did not. Three AD cases reverted to MCI status and 18 MCI cases reverted to control status. Removal of these individuals whose disease status reverted did not alter the presented results.

GENOTYPE DATA

Individual-level genotype data in the ADNI database¹⁵ were downloaded and merged to form a single data set containing genome-wide information for 818 individuals. Genetic analyses were performed using PLINK version 1.07 (<http://pngu.mgh.harvard.edu/~purcell/plink/>). Filtering criteria applied to individuals and SNPs are shown in the **Figure**.

Population structure was assessed by performing principal component analysis on a subset of all SNPs selected using multiple criteria (Figure). We assigned genotype-determined ancestry by comparing ADNI patients and reference populations from HapMap Phase 3 data. To control for population stratification, only individuals clustering with European HapMap samples were retained for analysis.

Quality control of genotype data for analyzed individuals included filters for missingness, heterozygosity, and concordance between genotype-determined and reported sex. The SNP quality control included filters for minor allele frequency (MAF), missingness, Hardy-Weinberg equilibrium, and differential missingness by case-control status. A total of 746 individuals passing quality control criteria were reclustered by performing principal component analysis.

MRI DATA

The ADNI MRIs were acquired at multiple sites using a GE Healthcare (Buckinghamshire, England), Siemens Medical Solutions USA (Atlanta, Georgia), or Philips Electronics 1.5 T system (Philips Electronics North America; Sunnyvale, California). Two high-resolution T1-weighted volumetric magnetization-prepared 180° radiofrequency pulses and rapid gradient-echo scans were collected for each study participant, and the raw Digital Imaging and Communications in Medicine images were downloaded from the public ADNI site (<http://www.loni.ucla.edu/ADNI/Data/index.shtml>). Parameter values can be found at <http://www.loni.ucla.edu/ADNI/Research/Cores/>.

All MRIs were processed according to previously published methods.⁸ Briefly, all MRIs were processed using the FreeSurfer version 4.1.0 software package (<http://surfer.nmr.mgh.harvard.edu>). A single magnetization-prepared 180° radiofrequency and rapid gradient-echo acquisition for each participant was normalized for intensity in homogeneities, nonbrain tissue removed, and subcortical white matter and deep gray matter volumetric structures segmented.^{16,17} Intensity gradients were followed outward from the white matter surface to find the gray matter surface (gray-cerebrospinal fluid boundary).^{18,19} Cortical thickness measurements were then obtained by calculating the distance between the gray and white matter surfaces at each point (per hemisphere) across the entire cortical surface.¹⁹ In our analyses, the cortical thickness and right-brain/left-brain volumes were averaged. To account for differences in head size, the total volume for each subcortical region of interest was corrected using a previously validated estimate of the total intracranial volume.^{19,20}

SNP SELECTION

Four GWAS-validated AD loci were selected for analysis: *APOE*, *CLU*, *PICALM*, and *CRI*. Genotypes of *APOE* were separately obtained via targeted genotyping, whereas SNPs showing the strongest degree of association in published GWASs^{4,5} were selected for analysis: rs11136000 at *CLU*, rs3851179 at *PICALM*, and rs1408077 at *CRI*. We selected for additional analysis all GWAS-promising SNPs with $P < 1 \times 10^{-3}$ in the prior GWASs.^{4,5} When multiple variants in moderate to high linkage disequilibrium at 1 locus ($r^2 > 0.6$) were reported to be associated with AD, only the SNP with the lowest P value was selected for analysis in the present study. Sixteen SNPs were chosen based on these criteria (**Table 1**).

NEUROIMAGING MEASURE SELECTION

Six neuroimaging measures were chosen for analysis on the basis of their established role in predicting AD risk and

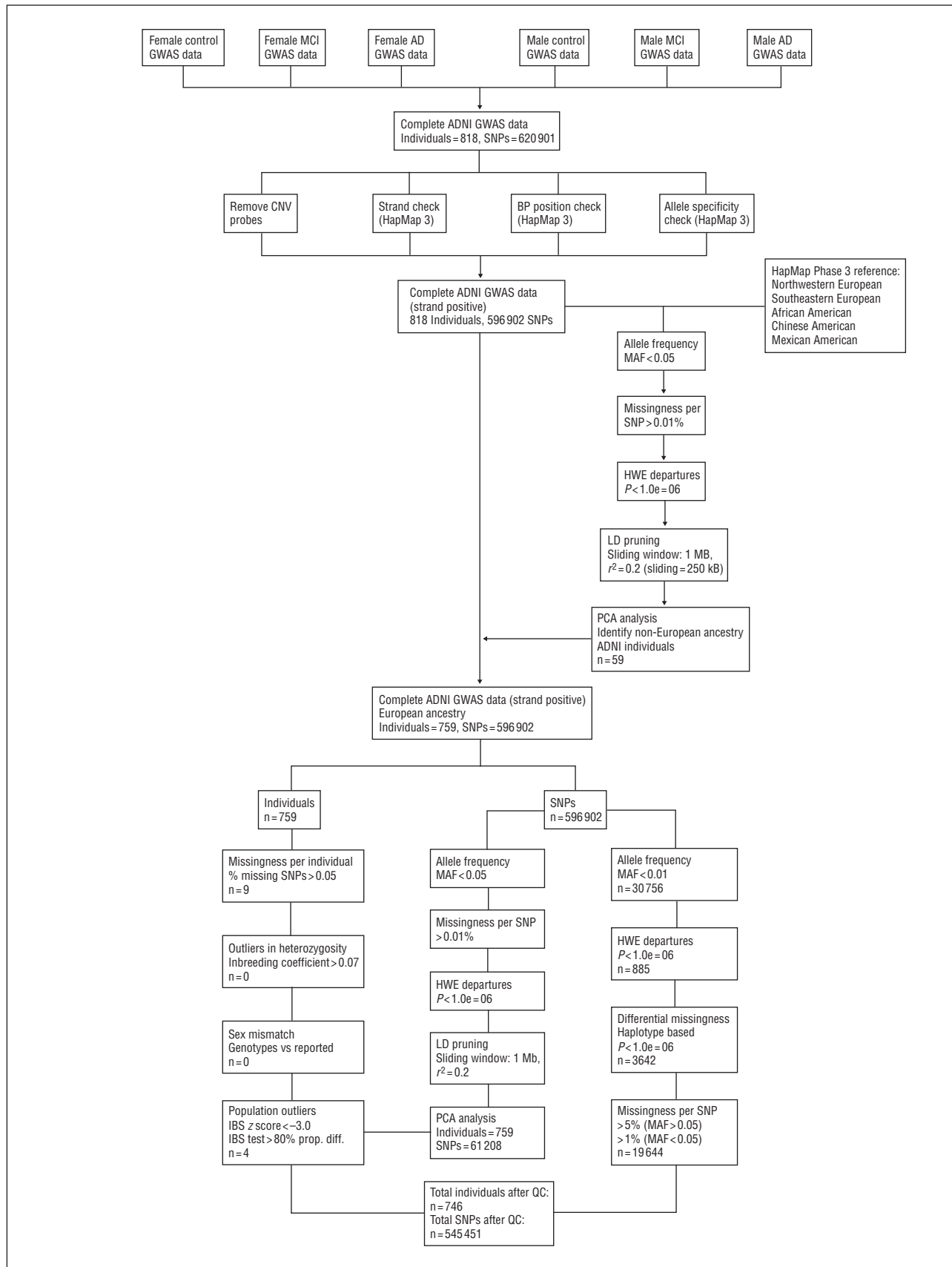


Figure. Genotype data quality control for the Alzheimer's Disease Neuroimaging Initiative (ADNI) genome-wide association study (GWAS) data set. Single-nucleotide polymorphism (SNPs) may have met multiple filtering criteria. AD indicates Alzheimer disease; BP, blood pressure; CNV, copy number variation; HWE, Hardy-Weinberg equilibrium; IBS, identity by state; LD, linkage disequilibrium; MAF, minor allele frequency; MCI, mild cognitive impairment; PCA, principal component analysis; prop. diff., proportional between-individuals difference as determined by identity by state; and QC, quality control.

Table 1. Prior GWASs and Current Study (Ordinal Logistic Regression) Results for Analyzed SNPs^a

SNP	Gene	GWAS OR	Included in Score?	Score Weight	Ordinal Logistic Regression (ADNI)	
					OR (95% CI) FDR-Corrected	P Value
rs11136000	<i>CLU</i>	0.84	Yes	-0.1744	0.97 (0.80-1.17)	.76
rs3851179	<i>PICALM</i>	0.85	Yes	-0.1625	0.99 (0.81-1.20)	.88
rs1408077	<i>CR1</i>	1.17	Yes	0.1570	1.27 (1.03-1.63)	.02
rs9384428	<i>ARID1B</i>	1.14	No	NA	0.93 (0.77-1.14)	.49
rs1539053	<i>DAB1</i>	0.88	No	NA	0.95 (0.76-1.16)	.51
rs1157242	<i>KCNU1</i>	1.17	No	NA	0.98 (0.75-1.25)	.89
rs676309	<i>MS4A4E</i>	1.14	No	NA	1.07 (0.88-1.31)	.50
rs662196	<i>MS4A6A</i>	0.88	No	NA	0.94 (0.76-1.14)	.48
rs7561528	<i>BIN1</i>	1.17	Yes	0.1570	1.29 (1.03-1.62)	.03
rs9446432	<i>C6orf155</i>	1.28	No	NA	0.97 (0.68-1.38)	.86
rs10501927	<i>CNTN5</i>	1.18	Yes	0.1655	1.25 (1.03-1.62)	.03
rs11894266	<i>SSB</i>	0.86	No	NA	0.97 (0.81-1.17)	.77
rs11952762	<i>DTWD2</i>	1.18	No	NA	1.12 (0.70-1.83)	.62
rs12201301	<i>C6orf205</i>	0.83	No	NA	1.02 (0.63-1.63)	.95
rs10499889	<i>SEMA3D</i>	1.15	No	NA	1.05 (0.87-1.27)	.63
rs8055533	<i>KIAA0350</i>	0.89	No	NA	1.02 (0.83-1.25)	.84

Abbreviations: ADNI, Alzheimer's Disease Neuroimaging Initiative; CI, confidence interval; FDR, false discovery rate; GWAS, genome-wide association study; NA, not applicable; OR, odds ratio; SNP, single-nucleotide polymorphism.

^aAll SNPs listed showed $P < .001$ in prior GWASs.^{4,5}

Table 2. Correlation Matrix for Analyzed Neuroimaging MRI Measures^a

	Hippocampal Volume	Amygdala Volume	WML Volume	Entorhinal Cortex Thickness	Parahippocampal Gyrus Thickness	Temporal Pole Thickness
Hippocampal volume		$P < .001$	$P < .001$	$P < .001$	$P < .001$	$P < .001$
Amygdala volume	0.7501		$P < .001$	$P < .001$	$P < .001$	$P < .001$
WML volume	-0.3830	-0.2312		$P < .001$	$P < .001$	$P < .001$
Entorhinal cortex thickness	0.6509	0.6348	-0.3083		$P < .001$	$P < .001$
Parahippocampal gyrus thickness	0.4866	0.4355	-0.3264	0.5401		$P < .001$
Temporal pole thickness	0.5330	0.5274	-0.2931	0.7183	0.4555	

Abbreviations: MRI, magnetic resonance imaging; WML, white matter lesion.

^aCorrelation coefficients (Spearman) and corresponding P values for comparison of analyzed MRI measures are shown.

progression: hippocampal volume, amygdala volume, white matter lesion (WML) volume, entorhinal cortex thickness (ECT), parahippocampal gyrus thickness, and temporal pole cortex thickness (TPT).²¹⁻²³ All analyzed neuroimaging measures were highly associated with AD in case-control analysis ($P < 1 \times 10^{-4}$ for all). However, correlation matrix analysis (**Table 2**) revealed limited association between measures (correlation coefficient range, -0.38 to 0.75), suggesting that independent analysis was needed for association with genetic variants.

GENETIC ASSOCIATION ANALYSIS

Genotype data were analyzed using an additive model, with odds ratios (ORs) or regression coefficients expressing the effect of each copy of the reference allele. Analyses of diagnostic categories (AD, MCI converters, MCI nonconverters, and controls) used an ordinal logistic regression model. Analyses of neuroimaging measures used linear regression. Continuous measures with skewed distributions were log transformed. All analyses included age, sex, history of hypertension, *APOE* genotype (number of $\epsilon 2$ and $\epsilon 4$ copies), alcohol abuse (*Diagnostic and Statistical Manual of Mental Disorders* [Fourth Edition]²⁶ criteria), and smoking status (ever smoker) as covariates. Education level was adjusted for according to number of school years

attended (<13, 13-16, or >16 years). Population stratification was adjusted for by incorporating the first 2 principal components as covariates. Neuroimaging analysis was performed independent of diagnostic category. Because Bonferroni correction was inappropriate owing to the nonindependence of tests, we used the false discovery rate (FDR) according to the method developed by Hochberg and Benjamini²⁷ to control for multiple hypothesis testing. Statistical significance was defined for FDR-corrected $P < .05$.

POWER CALCULATIONS FOR NEUROIMAGING ANALYSIS

We determined statistical power for identification of the association between analyzed variants and neuroimaging measures at a conservative $\alpha = .001$. To do so, we used the Genetic Power Calculator (<http://pngu.mgh.harvard.edu/~purcell/gpc/>).²⁸

SCORE-BASED ANALYSIS

Combined effects of non-*APOE* candidate SNPs were evaluated using a cumulative score-based method that was previously used to assess the cumulative effect of loci affecting lipid levels,²⁹ risk of myocardial infarction,³⁰ and blood pressure.³¹

Table 3. Baseline Demographic, Clinical, and Neuroimaging Characteristics of Study Participants^a

Characteristic	All Participants (N=740)	Cognitively Normal Controls (n=215)	MCI Nonconverters (n=217)	MCI Converters (n=140)	Patients With Alzheimer Disease (n=168)
Prospective follow-up time, median (IQR), mo	12 (6-24)	12 (6-24)	12 (6-24)	18 (6-24)	12 (6-24)
Age, mean (SD), y	75.3 (6.9)	75.9 (5.5)	75.3 (7.4)	74.6 (6.8)	75.5 (7.7)
Male sex, No. (%)	303 (40.9)	97 (45.1)	74 (34.1)	55 (39.3)	81 (48.2)
Education level, median (IQR), y	16 (14-18)	16 (14-18)	16 (13-18)	16 (14-18)	16 (12-17)
History of hypertension, No. (%)	348 (47.0)	97 (45.1)	102 (47.0)	69 (49.3)	86 (51.2)
Smoking, No. (%)	289 (39.0)	78 (36.3)	91 (41.9)	55 (39.3)	67 (39.9)
Alcohol abuse, No. (%)	30 (4.1)	5 (2.3)	8 (3.7)	7 (5.0)	10 (5.9)
<i>APOE</i> ε2, minor allele frequency	0.042	0.070	0.039	0.021	0.027
<i>APOE</i> ε4, minor allele frequency	0.367	0.142	0.290	0.432	0.430
GDS score, median (IQR)	1 (0-2)	0 (0-1)	1 (0-2)	1 (1-2)	1 (1-2)
ADAS-COG score, median (IQR)	10.7 (6.8-15.0)	6.0 (4.0-8.0)	10.7 (7.3-13.0)	13.0 (10.7-15.5)	18.0 (14.2-22.2)
Mini-Mental State Examination score, median (IQR)	27 (25-29)	29 (29-30)	28 (26-29)	27 (25-28)	23 (22-25)
Clinical Dementia Rating Sum of Boxes, median (IQR)	1.5 (0.0-3.0)	0.0 (0.0-0.0)	1.5 (1.0-2.0)	2.0 (1.0-2.5)	4.0 (3.5-5.0)
White matter lesion volume, median (IQR), cm ³	4.7 (3.2-7.4)	4.0 (2.9-5.9)	4.4 (3.1-7.5)	5.1 (3.7-7.0)	5.9 (4.1-10.2)
Amygdala volume, mean (SD), cm ³	1.25 (0.22)	1.38 (0.19)	1.28 (0.20)	1.16 (0.19)	1.11 (0.18)
Hippocampal volume, mean (SD), cm ³	3.23 (0.59)	3.66 (0.50)	3.28 (0.52)	2.95 (0.48)	2.85 (0.49)
Parahippocampal gyrus cortical thickness, mean (SD), mm	2.34 (0.32)	2.49 (0.28)	2.36 (0.33)	2.27 (0.27)	2.18 (0.32)
Temporal pole cortical thickness, mean (SD), cm ³	3.48 (0.13)	3.66 (0.26)	3.52 (0.32)	3.37 (0.36)	3.29 (0.39)
Entorhinal cortical thickness, mean (SD), mm	3.09 (0.47)	3.40 (0.30)	3.16 (0.45)	2.93 (0.43)	2.73 (0.43)

Abbreviations: ADAS-COG, Alzheimer Disease Assessment Scale Cognitive Subscale; GDS, Geriatric Depression Scale; IQR, interquartile range; MCI, mild cognitive impairment.

^aAll reported values (except follow-up time) refer to baseline ascertainment procedures. All volumetric measurements were adjusted to intracranial volume. Reported follow-up times were assessed on December 1, 2009.

Under this model each individual is assigned a score determined by multiplying the number of allele copies for SNPs of interest by a prespecified score weight. Score weights were based on β -coefficients extracted from case-control results from published GWAS reports^{4,5} (Table 1). Contributors to the genetic risk score included previously validated loci from GWASs (*CLU*, *PICALM*, and *CR1*) and those SNPs achieving adjusted significance (FDR-corrected $P < .05$) in our ordinal logistic regression analysis (*BIN1* and *CNTN5*). Score analysis performed without *BIN1* and *CNTN5* (data not shown) did not alter the results. Contributions from individual SNPs were summed to obtain a single genetic risk score, which was divided into quartiles for normalization. Single-SNP and score-based ordinal logistic regression results were analyzed using a maximum-likelihood method to compare predictive power for disease status.

RESULTS

GENETIC DATA QUALITY CONTROL

A total of 818 individuals enrolled had genotype data available for analysis. Of these, 72 were excluded by quality control filters (Figure), whereas our image-processing tools failed to produce good-quality results on the MRIs of 6 individuals. Therefore, we analyzed 740 individuals with genotype and MRI data that met filtering criteria (Table 3). Filtering of genome-wide data generated a final analyzed data set that included 545 451 SNPs (Figure). Population stratification was assessed by computing genomic inflation factors for all phenotypes (diagnosis and neuroimaging measurements): all values were lower than 1.005 after correction for principal components.

STATISTICAL POWER

We had more than 0.95 power for discovery of associations between *APOE* and neuroimaging traits (effect size, 5% of variance; MAF, 0.37). Power for discovery of associations with individual non-*APOE* loci was below 0.30 (effect size, 1% of variance; MAF range, 0.10-0.40). We therefore chose to pool genetic effects using a validated score-based model.²⁸⁻³⁰ Statistical power for the score-based analyses was approximately 0.80 (effect size, 3% of variance).

GENETIC RISK FACTORS FOR AD

We sought to extend known associations of *APOE*, *CLU*, *PICALM*, and *CR1* with AD using a logistic regression model across 4 diagnostic categories: disease-free controls, MCI nonconverters, MCI converters, and AD patients (Table 4). The strongest association with clinical diagnosis was shown by *APOE* (OR, 2.07; 95% confidence interval [CI], 1.67-2.56; FDR-corrected $P < 1 \times 10^{-6}$). Of the 3 previously confirmed non-*APOE* AD loci, only *CR1* was replicated in the ADNI data set, with SNP rs1408077 showing a significant association (OR, 1.27; 95% CI, 1.03-1.63; FDR-corrected $P = .02$). Among GWAS-promising SNPs with adjusted $P < 1 \times 10^{-5}$ in GWASs, 2 variants showed significant association in our analysis: rs10501927 at *CNTN5* (OR, 1.25; 95% CI, 1.02-1.53; FDR-corrected $P = .03$) and rs7561528 at *BIN1* (1.29; 1.03-1.62; FDR-corrected $P = .03$).

The genetic risk score included the following SNPs: rs11136000 (*CLU*), rs3851179 (*PICALM*), rs1408077

Table 4. Influence of Single SNP and a Cumulative Genetic Risk Score on Clinical Diagnosis^{a,b}

SNP	Gene	OR (95% CI)	P Value	FDR-Corrected P Value
<i>APOE</i> locus $\epsilon 4$	<i>APOE</i> ($\epsilon 4$)	2.07 (1.67-2.56)	$<1 \times 10^{-6}$	$<1 \times 10^{-6}$
Validated loci				
rs11136000	<i>CLU</i>	0.97 (0.80-1.17)	.75	.76
rs3851179	<i>PICALM</i>	0.99 (0.81-1.20)	.87	.88
rs1408077	<i>CR1</i>	1.27 (1.03-1.63)	.02	.02
Novel candidate loci				
rs10501927	<i>CNTN5</i>	1.25 (1.02-1.53)	.03	.03
rs7561528	<i>BIN1</i>	1.29 (1.03-1.62)	.03	.03
Genetic risk score (cumulative effect)				
Genetic risk score quartiles		1.14 (1.04-1.25)	.001	.001

Abbreviations: CI, confidence interval; FDR, false discovery rate; OR, odds ratio; SNP, single-nucleotide polymorphism.

^aSNPs were selected based on results of prior genome-wide association studies^{4,5} with $P < 1 \times 10^{-5}$. Results are not shown for 11 SNPs at novel candidate loci with $P > .05$. The genetic risk score includes all (5 of 16) SNPs outside the *APOE* locus achieving $P < .05$ in ordinal logistic regression. All analyses are adjusted for age, sex, history of hypertension, education level (<13 , 13-16, or >16 years), alcohol abuse, smoking (ever smoker status), and principal components 1 and 2. Analyses for SNPs outside the *APOE* locus were also adjusted for *APOE* genotypes (number of $\epsilon 2$ and $\epsilon 4$ copies).

^bClinical diagnosis defined as cognitively normal controls, mild cognitive impairment not converted to Alzheimer disease, mild cognitive impairment conversion to Alzheimer disease, and Alzheimer disease.

(*CR1*), rs10501927 (*CNTN5*), and rs7561528 (*BIN1*). Ordinal logistic regression revealed an association between risk score quartiles and diagnostic status (OR, 1.14; 95% CI, 1.04-1.25; FDR-corrected $P = .001$). Comparison of predictive performance between score-based analysis and individual SNP analyses favored the cumulative effects model ($P = .03$). To account for the possible heterogeneity in genetic and imaging risk profiles within the group whose cases did not convert to MCI, we repeated all analyses after removal of these individuals and observed similar results (data not shown).

GENETIC RISK FACTORS FOR MRI MEASURES

We investigated the influence of *APOE* genotype and genetic risk score profile on each MRI measure (**Table 5**). The *APOE* $\epsilon 4$ allele was strongly associated with all measures except WML volume ($P = .44$). Genetic risk score quartiles predicted increasing severity of all MRI measures (FDR-corrected $P = .04$). On analyzing score-contributing SNPs individually, we identified associations for the GWAS-validated SNPs rs1408077 at *CR1* with ECT (FDR-corrected $P = .03$) and rs3851179 at *PICALM* with hippocampal volume and ECT (FDR-corrected $P = .05$ and FDR-corrected $P = .01$, respectively). Furthermore, we identified associations for GWAS-promising SNPs rs10501927 at *CNTN5* with WML volume, parahippocampal gyrus thickness, TPT, and ECT (FDR-corrected $P = .002$, $P = .05$, $P = .02$, and $P = .02$, respectively) and for rs7561528 at *BIN1* with TPT and ECT (FDR-corrected $P = .03$ and $P = .01$, respectively).

COMMENT

Our results indicate that *APOE* and other previously validated loci for AD affect clinical diagnosis of AD and neuroimaging measures associated with disease. These findings suggest that sequence variants that modulate AD risk in recent GWASs may act through their influence on neuroimaging measures. Furthermore, our genetic analysis

of neuroimaging traits identified *BIN1* and *CNTN5* as genes of heightened interest for their relationship with AD, prioritizing these targets for further study.

Among non-*APOE* AD loci that have emerged from GWASs, only the *CR1* locus was significantly associated with disease status. Failure to extend previous findings for *CLU* and *PICALM* is likely because of the limited sample size of the ADNI cohort. Nonetheless, our genetic risk score was associated in a dose-dependent manner with clinical diagnosis and clearly outperformed individual SNP models. This finding is consistent with a biological role for at least some, if not all, of the incorporated loci. Interestingly, the inclusion of previously unvalidated loci at *BIN1* and *CNTN5* (albeit supported by $P < 1 \times 10^{-5}$ in the previous GWASs) did not degrade the performance of the genetic score, further supporting a role for these loci in AD.

The genetic risk score quartiles correlated with every examined neuroimaging trait, consistent with the underlying hypothesis that these traits are, at least in part, determined by genome sequence at these loci. This finding offers parallel evidence that the included genes influence biological processes underlying development of AD.

Among GWAS-validated loci, *APOE*, *PICALM*, and *CR1* genotypes influenced neuroimaging measures, whereas *CLU* did not. The robust effect of *APOE* was seen across all measures except WML volume, whereas the effect of *PICALM* was restricted to hippocampal volume and ECT, and the effect of *CR1* was restricted to ECT. These findings raise the possibility that the biological effects of these genes may be relatively confined to 1 neuroimaging trait and hence may offer clues to the mechanisms through which particular genetic variants might influence AD risk.

Two loci, identified as GWAS-promising in previous AD studies, showed association with neuroimaging measures. *CNTN5* variation was associated with WML, ECT, parahippocampal gyrus thickness, and TPT, whereas *BIN1* was associated with ECT and TPT. These genes encode proteins involved in neurite growth,³² presynaptic cytoskeleton structure integrity,³¹ and fission of synaptic vesicles.³³ Brain-specific isoforms and expression pat-

Table 5. Influence of Single SNPs and Cumulative Genetic Risk Score on Neuroimaging Measures^a

SNP	Gene	Coefficient (SE)	P Value	FDR-Corrected P Value
White matter lesion volume				
ε4	<i>APOE</i>	0.025 (0.033)	.44	.44
rs11136000	<i>CLU</i>	-0.030 (0.031)	.31	.32
rs3851179	<i>PICALM</i>	-0.005 (0.032)	.97	.98
rs1408077	<i>CR1</i>	0.028 (0.039)	.45	.46
rs10501927	<i>CNTN5</i>	0.119 (0.037)	.002	.002
rs7561528	<i>BIN1</i>	0.017 (0.032)	.50	.51
Genetic risk score quartiles		0.043 (0.015)	.04	.04
Hippocampal volume				
ε4	<i>APOE</i>	-0.240 (0.030)	0.9×10^{-14}	1.3×10^{-14}
rs11136000	<i>CLU</i>	-0.019 (0.030)	.78	.79
rs3851179	<i>PICALM</i>	0.061 (0.029)	.04	.05
rs1408077	<i>CR1</i>	-0.037 (0.038)	.32	.32
rs10501927	<i>CNTN5</i>	-0.046 (0.036)	.17	.19
rs7561528	<i>BIN1</i>	-0.055 (0.031)	.06	.08
Genetic risk score quartiles		-0.099 (0.014)	.001	.002
Amygdala volume				
ε4	<i>APOE</i>	-0.079 (0.012)	3.6×10^{-11}	3.9×10^{-11}
rs11136000	<i>CLU</i>	-0.018 (0.012)	.11	.12
rs3851179	<i>PICALM</i>	0.009 (0.012)	.47	.47
rs1408077	<i>CR1</i>	-0.017 (0.014)	.21	.22
rs10501927	<i>CNTN5</i>	-0.018 (0.013)	.19	.19
rs7561528	<i>BIN1</i>	-0.020 (0.012)	.09	.10
Genetic risk score quartiles		0.043 (0.015)	.02	.02
Entorhinal cortex thickness				
ε4	<i>APOE</i>	-0.127 (0.026)	8.7×10^{-7}	9.1×10^{-7}
rs11136000	<i>CLU</i>	-0.011 (0.025)	.65	.67
rs3851179	<i>PICALM</i>	0.066 (0.021)	.01	.01
rs1408077	<i>CR1</i>	-0.067 (0.031)	.03	.03
rs10501927	<i>CNTN5</i>	-0.067 (0.025)	.02	.02
rs7561528	<i>BIN1</i>	-0.121 (0.025)	.004	.01
Genetic risk score quartiles		-0.048 (0.011)	7.9×10^{-4}	8.4×10^{-4}
Parahippocampal gyrus cortex thickness				
ε4	<i>APOE</i>	-0.063 (0.017)	3.3×10^{-4}	3.8×10^{-4}
rs11136000	<i>CLU</i>	0.007 (0.017)	.66	.67
rs3851179	<i>PICALM</i>	0.014 (0.017)	.29	.30
rs1408077	<i>CR1</i>	0.0004 (0.021)	.98	.98
rs10501927	<i>CNTN5</i>	-0.040 (0.019)	.05	.05
rs7561528	<i>BIN1</i>	-0.019 (0.017)	.24	.24
Genetic risk score quartiles		-0.022 (0.010)	.04	.04
Temporal pole cortex thickness				
ε4	<i>APOE</i>	-0.061 (0.019)	.002	.004
rs11136000	<i>CLU</i>	-0.011 (0.018)	.50	.51
rs3851179	<i>PICALM</i>	0.033 (0.017)	.06	.06
rs1408077	<i>CR1</i>	-0.031 (0.024)	.12	.14
rs10501927	<i>CNTN5</i>	-0.051 (0.022)	.02	.02
rs7561528	<i>BIN1</i>	-0.041 (0.019)	.02	.03
Genetic risk score quartiles		-0.025 (0.009)	8.2×10^{-4}	.001

Abbreviations: FDR, false discovery rate; SNP, single-nucleotide polymorphism.

^aSNPs were selected based on results of prior genome-wide association studies^{4,5} with $P < .001$. Results are not shown for 11 SNPs at novel candidate loci with $P > .05$. The genetic risk score includes all (5 of 16) SNPs outside the *APOE* locus achieving $P < .05$ in ordinal logistic regression. All analyses are adjusted for age, sex, history of hypertension, education level (<13, 13-16, or >16 years), alcohol abuse, smoking (ever smoker status), and principal components 1 and 2. Analyses for SNPs outside the *APOE* locus were also adjusted for *APOE* genotypes (number of ε2 and ε4 copies).

terns have been reported for *BIN*³⁴ and *CNTN5*.³⁵ Although our results for these loci can only be considered preliminary, they may help prioritize targets for future genetic studies and GWASs in AD, particularly given their association with neuroimaging correlates of AD and disease status.

The crucial limitations of our study arise from its small sample size. Because of restricted power, we were forced to constrain our analysis to SNPs and loci with high prior

probabilities of association with AD and imaging traits, based on their status as either validated (*APOE*, *CLU*, *PICALM*, and *CR1*) or promising (*CNTN5* and *BIN1*) genetic risk factors. Our power also limits the conclusions we can draw about observed differential genetic effects on neuroimaging traits. For example, although the absence of an effect of *CR1* on hippocampal volume may reflect important biology, it is also possible that an effect could be detected with increased power.

In summary, we have shown that established and candidate AD genes have a role in 6 neuroimaging traits linked to AD. Furthermore, 2 promising genes from prior AD GWASs, *CNTN5* and *BIN1*, are also associated with these neuroimaging measures, which heightens their interest as novel AD loci. These genes may act selectively, influencing only 1 or a few established AD-related MRI measures. Future studies are required to replicate and expand these findings.

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Correspondence: Jonathan Rosand, MD, MSc, Center for Human Genetic Research, Department of Neurology, Massachusetts General Hospital, 185 Cambridge St, Ste CPZN-6818, Boston, MA 02114 (jrosand@partners.org).

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