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Polygenic Hazard Scores in Preclinical Alzheimer Disease

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Identifying asymptomatic older individuals at elevated risk for developing Alzheimer disease (AD) is of clinical importance. Among 1,081 asymptomatic older adults, a recently validated polygenic hazard score (PHS) significantly predicted time to AD dementia and steeper longitudinal cognitive decline, even after controlling for APOE E4 carrier status. Older individuals in the highest PHS percentiles showed the highest AD incidence rates. PHS predicted longitudinal clinical decline among older individuals with moderate to high Consortium to Establish a Registry for Alzheimer’s Disease (amyloid) and Braak (tau) scores at autopsy, even among APOE ε4 noncarriers. Beyond APOE, PHS may help identify asymptomatic individuals at highest risk for developing Alzheimer neurodegeneration.

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Beyond APOE ε4 carrier status, recent genetic studies have identified numerous single nucleotide polymorphisms (SNPs), each of which is associated with a small increase in AD dementia risk.6 Using genome-wide association studies (GWAS) from AD cases and controls, we have recently developed a novel polygenic hazard score (PHS) for predicting age-specific risk for AD dementia that integrates 31 AD-associated SNPs (in addition to APOE) with U.S. population-based AD dementia incidence rates.7 Among asymptomatic older adults, in retrospective analyses, we have previously shown that the PHS-predicted age of AD onset strongly correlates with the actual age of onset.7 To evaluate clinical usefulness and further validate PHS, in this study, we prospectively evaluated whether PHS predicts rate of progression to AD dementia and longitudinal cognitive decline in cognitively asymptomatic older adults and individuals with MCI.

Subjects and Methods

We evaluated longitudinal clinical and neuropsychological data (from March 2016) from the National Alzheimer’s Coordinating Center (NACC).8 Using the NACC uniform dataset, we focused on older individuals classified at baseline as cognitively normal (CN), with a Clinical Dementia Rating (CDR) of 0 and available genetic information (n = 1,081, Table). We also...
evaluated older individuals classified at baseline as MCI (CDR = 0.5) with available genetic information (n = 571, see Table). We focused on CN and MCI individuals with age of AD dementia onset < 88 years to avoid violations of Cox proportional hazards assumption as evaluated using scaled Schoenfeld residuals (total n = 1,652). The institutional review boards of all participating institutions approved the procedures for all Alzheimer’s Disease Genetics Consortium (ADGC) and National Institute on Aging Alzheimer’s Disease Center substudies. Written informed consent was obtained from all participants or surrogates.

For each CN and MCI participant in this study, we calculated their individual PHS, as previously described.\(^7\) In brief, we identified AD-associated SNPs (at \(p < 10^{-5}\)) using genotype data from 17,008 AD cases and 37,154 controls from Stage 1 of the International Genomics of Alzheimer’s Disease Project. Next, we selected a final total of 31 SNPs based on a stepwise Cox proportional hazards model using genotype data from 6,409 AD patients and 9,386 older controls from Phase 1 of the ADGC. We corrected the resulting scores for each individual for the baseline allele frequencies using European genotypes from 1000 Genomes Project to derive a PHS for each individual. Finally, by combining U.S. population-based incidence rates and genotype-derived PHS for each individual, we derived estimates of instantaneous risk for developing AD, based on genotype and age. In this study, the PHS computed for every CN and MCI participant represents the vector product of an individual’s genotype for the 31 SNPs and the corresponding parameter estimates from the ADGC Phase 1 Cox proportional hazard model in addition to the \(APOE\) effects (for additional details see Desikan et al\(^7\)).

We first investigated the effects of the PHS on progression to AD dementia by using a Cox proportional hazards model, with time to event indicated by age of AD dementia onset. We resolved “ties” using the Breslow method. We covaried for effects of sex, \(APOE\) \(\varepsilon4\) status (binarized as having at least 1 \(\varepsilon4\) allele vs none), education, and age at baseline. To prevent violating the proportional hazards assumptions, we additionally included baseline age stratified by quintiles as a covariate.\(^10\)

Next, we employed a linear mixed-effects (LME) model to evaluate the relationship between PHS and longitudinal clinical decline as assessed by change in CDR Sum of Boxes (CDR-SB) as well as by change in Logical Memory Test, Wechsler Adult Intelligence Scale–Revised (WAIS-R) Digit Symbol, Boston Naming Test (BNT), Trail-Making Tests A and B (TMTA/B), and forward and backward Digit Span Test (f/bDST). To maintain consistent directionality across all tests, we inverted the scale for Trail-Making tests such that lower scores represent decline. We covaried for sex, \(APOE\) \(\varepsilon4\) status, education, baseline age, and all their respective interactions with time. We were specifically interested in the PHS \(\times \) time interaction, whereby a significant interaction indicates differences in rates of decline, as a function of differences in PHS. We then examined the main effect of PHS by comparing slopes of cognitive decline over time in the neuropsychological tests for individuals with high (~84 percentile) and low PHS (~16 percentile), defined by 1 standard deviation above or below the mean of PHS, respectively.\(^11\) We also compared goodness of fit between the LME models with and without PHS using likelihood ratio tests to determine whether PHS LME models resulted in a better fit than non-PHS LME models.

Finally, we evaluated the relationship between PHS, \(APOE\), and neuropathology in preclinical AD. Specifically, we conducted LME analysis assessing longitudinal change in CDR-SB in CN individuals with available neuropathology (specifically, neuritic plaque scores based on the Consortium to Establish a Registry for Alzheimer’s Disease [CERAD] and neurofibrillary tangle scores assessed with Braak stages).

### Results

PHS significantly predicted risk of progression from CN to AD dementia (hazard ratio [HR] = 2.36, 95% confidence interval [CI] = 1.38–4.03, \(p = 1.66 \times 10^{-3}\)), illustrating that polygenic information beyond \(APOE\) \(\varepsilon4\) can identify asymptomatic older individuals at greatest risk for developing AD dementia. Individuals in the highest PHS decile had the highest annualized AD

### Table . Cohort Demographics

<table>
<thead>
<tr>
<th>Characteristic</th>
<th>CN, n = 1,081</th>
<th>MCI, n = 571</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age, yr (± SD)</td>
<td>71.19 (6.65)</td>
<td>74.70 (5.85)</td>
</tr>
<tr>
<td>Education, yr (± SD)</td>
<td>16.07 (2.57)</td>
<td>15.70 (2.91)</td>
</tr>
<tr>
<td>F, No. [%]</td>
<td>719 [66.51]</td>
<td>291 [50.96]</td>
</tr>
<tr>
<td>(APOE) (\varepsilon4) carriers, No. [%]</td>
<td>297 [27.47]</td>
<td>347 [60.77]</td>
</tr>
<tr>
<td>Converted to AD dementia, No. [%]</td>
<td>38 [3.52]</td>
<td>390 [68.30]</td>
</tr>
<tr>
<td>Baseline MMSE (± SD)</td>
<td>29.22 (1.05)</td>
<td>25.67 (3.36)</td>
</tr>
</tbody>
</table>

AD = Alzheimer disease; CN = cognitively normal; F = female; MCI = mild cognitive impairment; MMSE = Mini-Mental State Examination; SD = standard deviation.
Using the combined CN and MCI cohorts (total n = 5,652) to maximize statistical power, we found that PHS predicted risk of progression from CN and MCI to AD dementia (HR = 1.17, 95% CI = 1.02–1.35, p = 2.36 × 10^{-2}). Using the combined CN and MCI cohorts (total n = 5,652) to maximize statistical power, we found that PHS significantly predicted risk of progression from CN and MCI to AD dementia (HR = 1.31, 95% CI = 1.14–1.51, p = 1.82 × 10^{-4}). At 50% risk for progressing to AD dementia, the expected age for developing AD dementia is approximately 85 years for an individual with low PHS (~16 percentile); however, for an individual with high PHS (~84 percentile), the expected age of onset is approximately 78 years. In all Cox models, the proportional hazard assumption was valid for all covariates.

Evaluating clinical progression and cognitive decline within the CN individuals, we found significant PHS × time interactions for CDR-SB (β = 0.05, standard error [SE] = 0.02, p = 3.64 × 10^{-4}), WAIS-R (β = −0.61, SE = 0.30, p = 4.25 × 10^{-2}), TMTB (β = −2.48, SE = 0.99, p = 1.20 × 10^{-3}), and fDST (β = −0.93, SE = 0.45, p = 3.76 × 10^{-2}; Supplementary Table 1), with significantly steeper slopes for high PHS individuals for WAIS-R, TMTB, and CDR-SB (Supplementary Table 2, Fig 2). Evaluating average percentage change across all neuropsychological tests, we found that PHS predicted cognitive decline (β = 0.84, SE = 0.30, p = 4.50 × 10^{-5}), with high PHS individuals showing greater rates of decline (β = −1.80, SE = 0.89, p = 4.30 × 10^{-5}) compared to low PHS individuals (β = −0.12, SE = 0.80, p = 0.88). Goodness of fit comparison using likelihood ratio tests showed that the full LME model comprising PHS and covariates resulted in a better model fit for predicting decline in CDR-SB, BNT, WAIS-R, fDST, and TMTB (Supplementary Table 3). We found similar results within the MCI individuals and the combined CN and MCI cohort (Supplementary Tables 1–7), illustrating that polygenic information beyond APOE e4 can identify asymptomatic and mildly symptomatic individuals at highest risk for clinical and cognitive decline.

Finally, among CN individuals with moderate and frequent CERAD C score at autopsy, we found that PHS predicted change in CDR-SB over time (β = 1.25, SE = 0.28, p = 6.63 × 10^{-6}), with high PHS individuals showing a greater rate of increase (β = 5.62, SE = 0.92, p = 1.23 × 10^{-3}). In a reduced model without PHS, APOE e4 status did not predict change in CDR-SB (β = 0.26, SE = 0.50, p = 0.61). Furthermore, even in APOE e4 noncarriers, PHS predicted change in CDR-SB over time (β = 2.11, SE = 0.38, p = 3.06 × 10^{-8}), with high PHS individuals showing a greater rate of increase (β = 6.11, SE = 1.08, p = 1.60 × 10^{-8}). Similarly, among CN individuals with Braak stage III–IV at autopsy, PHS predicted change in CDR-SB over time (β = 0.93, SE = 0.24, p = 1.11 × 10^{-4}), with high PHS individuals showing a greater rate of increase (β = 3.98, SE = 0.79, p = 4.45 × 10^{-7}).

Discussion

Here, we show that PHS predicts time to progress to AD dementia and longitudinal cognitive decline in both
preclinical AD and MCI. Among CN individuals with moderate to high CERAD and Braak scores at autopsy, we found that PHS predicted longitudinal clinical decline, even among APOE ε4 noncarriers. Beyond APOE, our findings indicate that PHS can be useful to identify asymptomatic older individuals at greatest risk for developing AD neurodegeneration.

These results illustrate the value of leveraging the polygenic architecture of the AD process. Building on prior work,4,5 our findings indicate that polygenic information may be more informative than APOE for predicting clinical and cognitive progression in preclinical AD. Although prior studies have used polygenic risk scores in preclinical AD,12–14 by focusing on maximizing differences between cases and controls, this approach is clinically suboptimal for assessing an age-dependent process like AD dementia, where a subset of controls will develop dementia over time (see Fig 1). Furthermore, given the bias for selecting diseased cases and normal controls, baseline hazard/risk estimates derived from GWAS samples cannot be applied to older individuals from the general population.15 By employing an age-dependent, survival analysis framework and integrating AD-associated SNPs with established population-based incidence rates,16 PHS provides an accurate estimate of age of onset risk in preclinical AD.

In our neuropathology analyses, PHS predicted longitudinal clinical decline in older individuals, with moderate to high amyloid or tau pathology indicating that PHS may serve as an enrichment strategy for secondary prevention trials. Congruent with recent findings that the risk of dementia among APOE ε4/4 is lower than previously estimated,17 among CN subjects with moderate to high amyloid load, we found that APOE did not predict clinical decline and PHS predicted change in CDR-SB even among APOE ε4 noncarriers. Our combined findings suggest that beyond APOE, PHS may prove useful as both a risk stratification and an enrichment marker to identify asymptomatic individuals most likely to develop Alzheimer neurodegeneration.

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Author Contributions


Potential Conflicts of Interest

Nothing to report.

References


