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ABCA7 p.G215S as potential protective factor for Alzheimer’s disease


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ABSTRACT

Genome-wide association studies (GWASs) have been effective approaches to dissect common genetic variability underlying complex diseases in a systematic and unbiased way. Recently, GWASs have led to the discovery of over 20 susceptibility loci for Alzheimer’s disease (AD).

Despite the evidence showing the contribution of these loci to AD pathogenesis, their genetic architecture has not been extensively investigated, leaving the possibility that low frequency and rare coding variants may also occur and contribute to the risk of disease. We have used exome and genome sequencing data to analyse the single independent and joint effect of rare and low frequency protein coding variants in 9 AD GWAS loci with the strongest effect sizes after APOE (BIN1, CLU, CR1, PICALM, MS4A6A, ABCA7, EPHA1, CD33, CD2AP) in a cohort of 332 sporadic AD cases and 676 elderly controls of British and North American ancestry. We identified coding variability in ABCA7 as contributing to AD risk. This locus harbors a low frequency coding variant (p.G215S, rs72973581, MAF=4.3%) conferring a modest but statistically significant protection against AD (p-value= 6x10^{-4}, OR= 0.57, 95% CI 0.41-0.80). Notably, our results are not driven by an enrichment of loss of function variants in ABCA7, recently reported as main pathogenic factor underlying AD risk at this locus. In summary, our study confirms the role of ABCA7 in AD and provide new insights that should address functional studies.

KEYWORDS
Alzheimer’s disease (AD); Genome-wide association studies (GWASs); ABCA7; whole exome sequencing (WES); whole genome sequencing (WGS); protective variant

1. INTRODUCTION

Alzheimer’s disease (AD) is the most common cause of progressive dementia in the elderly. Aging and genetic factors play a critical role for the disease development. Rare coding and fully penetrant mutations in APP, PSEN1 and PSEN2 explain part of the AD autosomal dominant cases. On the other hand, APOE ε4 allele and rare coding variants in TREM2
represent the main risk factors for late-onset and apparently sporadic AD (Chartier-Harlin et al., 1994) (Guerreiro et al., 2013). In the last 5 years, genome-wide association studies (GWASs) identified over 20 main risk loci influencing AD susceptibility (Harold et al., 2009)(Hollingworth et al., 2011)(Lambert et al., 2009)(Naj et al., 2011)(Seshadri et al., 2010)(Lambert et al., 2013). Among these, 9 have been replicated by at least two independent GWASs and present the strongest effect sizes after APOE (BIN1, CLU, CR1, PICALM, MS4A6A, ABCA7, EPHA1, CD33, CD2AP).

GWASs have been a successful strategy to identify loci associated to a common trait, shedding light on disease pathways, and for AD these include 1) immune response (CR1, MSA4A/MSA7A, CD2AP, CD33, EPHA1, ABCA7), 2) vesicles trafficking (PICALM, BIN1), 3) lipid metabolism (CLU, ABCA7) and 4) Aβ peripheral clearance (PICALM, BIN1, CD33 and ABCA7). (http://www.alzgene.org/). Nevertheless, the functional variant(s) within these risk loci have not yet been fully defined.

GWAS arrays tag common, low penetrant and generally non coding variants that likely exert a subtle regulatory effect (0.8<OR<1.5) on a trait, affecting gene expression, CpG islands methylation and splicing, in cis or trans (Visscher et al., 2012)(Ramasamy et al., 2014). Whereas, low frequency (1%<MAF<5%) and rare variants (MAF<1%) with a modest penetrance remain mostly undetected either because they are not in the array or because, even with the implementation of imputation, the detection of variants with MAF <2% is not sufficiently accurate. As an illustrative example, APOE GWAS hit maps to an intronic region and it is likely driven by the APOE e4 allele, which is a common coding haplotype (rs429358, p.C130R and rs7412, p.R176R MAF=15%) in exon 4, that it is not tagged by the custom genotyping arrays mostly used.

Recently, resequencing studies have been powerful strategies to bridge the gap between susceptibility loci identified and actual disease-modifying variant(s) (Beaudoin et al., 2013)(Rivas et al., 2011)(Service et al., 2014)(Lohmueller et al., 2013).

Therefore, we have used exome and genome sequencing data 1) to identify rare and low frequency coding variants in BIN1, CLU, CR1, PICALM, MS4A6A, ABCA7, EPHA1, CD33, CD2AP and 2) to investigate their single independent and combined effect on AD susceptibility. Both the single-variant and the gene-based association tests confirmed ABCA7 as
susceptibility locus associated with AD. Importantly, although $ABCA7$ loss of function (LoF) mutations (indels, nonsense, splice-site mutations) have been recently reported as main mechanism increasing AD risk at this locus (Steinberg et al., 2015), our results are not driven by such variants. Whereas, we report an enrichment for $ABCA7$ common and low frequency coding variants with a potential protective effect, that is mainly responsible for our gene-based signal. Among these, $ABCA7$ p.G215S is the main low frequency missense hit in the single-variant analysis in the discovery cohort. The potential protective role of this variant has been further confirmed in an independent European and North American cohort. Our results show that $ABCA7$ p.G215S exerts a mild but statistically significant influence, lowering the risk for AD. Thus, confirming $ABCA7$ to be a good potential target to address functional studies.

2. MATERIALS AND METHODS

The discovery cohort was composed of 332 apparently sporadic AD cases and 676 elderly controls, neuropathologically and clinically confirmed, originating from the UK and North America. The mean age at disease onset was 71.66 years (range 41-94 years) for cases and the mean age of ascertainment was 78.15 years (range 60-102 years) for controls (Table 1). The majority of the cases (77%) were late-onset (> 65 years at onset).

Among the cases and controls, 42% and 51% were female, respectively. 58% and 47% of the cases and controls carried the $APOE$ ε4 allele, respectively. The $APOE$ ε4 allele was significantly associated to the disease status in the NIH and ADNI series (p-value= 0.02 and $1.19\times10^{-9}$, respectively). The threshold call rate for inclusion of the subject in analysis was 95%. On this cohort we performed 1) gene-based analysis (SKAT and c-alpha tests) and 2) single-variant association analysis, targeting 23.5 Kilobase pairs (Kbs) of coding sequence. Finally, we followed-up, in an independent Caucasian dataset, $ABCA7$ p. G215S, the only nominal significant low frequency missense variant in the single-marker association test in our discovery set (Figure 1).

The follow-up dataset was composed of 307 late-onset apparently sporadic AD cases from North America and Europe and 501 elderly Caucasian controls from North America (Coriell repositories), Europe, Australia and Canada (Table 1). Written informed consent was
obtained for each clinically assessed individual and the study was approved by the appropriate institutional review boards. All samples had fully informed consent for retrieval and were authorized for ethically approved scientific investigation (UCLH Research Ethics Committee number 10/H0716/3, BYU IRB, Cardiff REC for Wales 08/MRE09/38+5, REC Reference 04/Q2404/130, National Research Ethics Service [NRES]).

2.1 Exome sequencing
DNA was extracted from blood or brain for cases and brain only for controls using standard protocols. Library preparation for next generation sequencing used DNA (between 1 µg and 3 µg) fragmented in a Covaris E210 (Covaris Inc.). DNA was end-repaired by 5’phosphorylation, using the Klenow polymerase. A poly-adenine tail was added to the 3’end of the phosphorylated fragment and ligated to Illumina adapters. After purification using an AMPure DNA Purification kit (Beckman Coulter, Inc), adapter-ligated products were amplified. The DNA library was then hybridized to an exome capture library (NimbleGen SeqCap EZ Exome v2.0, Roche NimbleGen Inc. or TruSeq, Illumina Inc.) and precipitated using streptavidin-coated magnetic beads (Dynal Magnetic Beads, Invitrogen). These exome libraries were PCR-amplified, and then DNA hybridized to paired-end flow cells using a cBot (Illumina, Inc.) cluster generation system. Samples were sequenced on the Illumina HiSeq™ 2000 using 2x100 paired end reads cycles.

2.2 Whole Genome sequencing
Genome sequencing was performed in 199 controls, from the Cache County Study on Memory in Aging. All samples were sequenced with the use of Illumina HiSeq technology.

2.3 Sanger sequencing
ABCA7 p. G215S (rs72973581) was screened in an additional follow-up cohort composed of 307 late-onset cases and 501 elderly controls. Primers were designed in Primer3 (http://bioinfo.ut.ee/primer3-0.4.0/) using UCSC (http://genome.ucsc.edu/) reference sequences NM_019112 (ABCA7).
Purified sequences were analysed on an ABI 3730 DNA Analyzer (Applied Biosystems, CA, USA) and chromatograms were visualized in Sequencher software (version 4.2 Gene Codes Corporation, MI, USA).

2.4 Bioinformatics

Sequence alignment and variant calling were performed against the reference human genome (UCSC hg19). Alignment was performed with the use of CASAVA software and variant calling was performed with the use of SAMtools (Li et al., 2009) and GATK (McKenna et al., 2010). Paired end sequence reads (2x100bp paired end read cycles) were aligned using the Burrows-Wheeler aligner (Li and Durbin, 2009). Format conversion and indexing were performed with Picard (www.picard.sourceforge.net/index.shtml). GATK was used to recalibrate base quality scores, perform local re-alignments around indels and to call and filter the variants (McKenna et al., 2010). VCFtools was used to annotate gene information for the remaining novel variants. We used ANNOVAR software to annotate the variants (Wang et al., 2010). Variants were checked against established databases (1000 Genomes Project and dbSNP v.134). Calling algorithms, pipe-lines and reference panels were the same the pooled datasets. The protein coding effects of variants was predicted using SIFT, Polyphen2 and SeattleSeq Annotation (gvs.gs.washington.edu/SeattleSeqAnnotation). All variants within the coding regions of the 9 risk loci (ABCA7 [NM_019112]; CD2AP [NM_012120]; MS4A6A [NM_152851]; CR1 [NM_000573]; BIN1 [NM_139343]; PICALM [NM_001206946]; EPHA1 [NM_005232]; CLU [NM_001831]; CD33 [NM_001772]) have been collected and analysed. Indels were excluded from the merged dataset because not targeted in the ADNI subcohort (Figure 1). (Further details are provided in the supplementary materials).
2.5 Statistical Analysis

In the single-variant analysis, allele frequencies were calculated for each low frequency and rare coding variants in cases and controls and Fisher’s exact test on allelic association was performed. To study the joint effect of the variants detected, we performed a gene-based analysis with SKAT and c-alpha test and we analysed together for each gene the whole spectrum of allelic variability (common, low frequency, rare, coding and non-coding).

C-alpha test and SKAT are closely related, being both non-burden test, analyzing and collapsing the effect of genetic variants of different frequency (common and rare), effect (protective, damaging and neutral) and effect size (modest, moderate, strong). SKAT can be considered an expansion of the c-alpha test because overcomes some of its limits. Indeed, SKAT 1) can be applied also to the study of continuous traits and 2) do not need any permutation.

Low frequency and rare variants were defined as having a 1%<MAF<5% and MAF<1%, respectively, either in cases or controls.

All computations, C-alpha and SKAT tests were performed in R (version x64 3.0.2, http://www.r-project.org/) and PLINK/SEQ.

A p-value of 0.05 was set as a nominal significance threshold. Based on multiple testing correction, the thresholds for single variant and gene-based association tests are defined by p-value=1.25x10^{-3} (0.05/40 [total number of coding low frequency and rare variants detected in our study]) and 5.5x10^{-3} (0.05/9genes), respectively. Furthermore, we excluded singletons from the single-variant analysis, since a variant observed only once is not largely informative about the overall distribution (Neale et al., 2011). However, we pooled the singletons together and analysed their collective effect in the gene-based analysis (SKAT and c-alpha test).

In addition, we report the complete list of coding variants detected in these GWAS loci in the supplementary table (Table S1).
3. RESULTS

The discovery set consisted of a total of 332 sporadic and mainly late-onset AD cases and 676 elderly controls of British and North American ancestry (Table 1).

A total of 289 single nucleotide variants (SNVs) were identified. Among these, 128 (44.3%) were nonsynonymous, 72 (24.9%) were synonymous, 83 (28.7%) UTR and 6 (2%) were intronic variants. Among the missense variants, 99 (77.34%) were rare (MAF<1%), and 72.72% of these (72) were singletons (a variant observed only once either in cases or controls). 15 nonsynonymous variants (11.7%) were low frequency (1 %< MAF<5%) and 16 (12.5%) were common (MAF≥5%). In addition, we report 14 novel coding variant (not reported in ExAC, released 13 January 2015, or dbSNP 137). None of the detected low frequency and rare coding variants cluster within common haplotypes (MAF≥5%) and therefore could have been missed by GWASs and chip based fine-mapping approaches. (Table S1, Table S2). We report the presence of two or three low frequency and/or rare variants in the studied genes in the same individual, both for cases and controls (Table S3a and Table S3b).

Overall, the total variant frequency of the 9 GWAS loci in our study was in line with the one reported for the European-American cohort in the Exome Variant Server (EVS)(http://evs.gs.washington.edu/EVS/). The only exception was represented by CR1, that showed a 2.7-fold higher relative frequency of total variants, compared to the EVS database. (Table S4).

PICALM harbors the lowest burden of low frequency and rare coding variants (3.27 coding variants per kb of coding sequence). By contrast, CD33, presents the highest relative frequency of coding variants and the lowest relative frequency of damaging variants (9.14 and 0.91 coding and damaging variants, respectively), suggesting that the majority of coding variability in CD33 is likely non-functional (Table S5).

BIN and ABCA7 display the highest relative proportion of damaging variants (3.92 [87.3% of its coding variability] and 3.72 [60% of its coding variability] damaging coding variants per kb of coding sequence, respectively), thus arguing for a potential functional impact of
missense mutations at these loci (Table S5). Moreover, ABCA7 was the only gene harboring nonsense mutations.

Most of the low frequency and rare coding variability identified within these loci exerts generally a relatively modest effect (mean OR=1.1) that is comparable to those observed for common and generally noncoding variants identified by GWASs (Table S1).

3.1 Single coding variant association test

The main hits of the single-variants association test map mainly to ABCA7 (Table 2). Particularly, we report ABCA7 p.G215S (rs72973581), that was the only low-frequency (MAF=4.3%) missense variant showing a trend towards significance in the single-marker association test (p-value =0.02 and corrected p-value=0.8) in the discovery set and was statistically significant after Bonferroni correction (p-value = 6x10^{-4} and corrected p-value=0.024) in the combined datasets (discovery set and follow-up dataset).

Rs72973581 [A] results in a glycine to serine amino acid change at the position 215 of ABCA7 (G215S) and its frequency was 1.56-fold higher in controls compared to cases (MAF=4.66 and 7.24 for cases and controls, respectively), arguing for a protective effect (OR=0.6, 95% CI=0.38-0.95). This variant was present in homozygosity in one control. The study possessed relatively low power to detect a significant association between cases and controls for low frequency and rare variants. Therefore, we have followed up ABCA7 p.G215S, carrying out Sanger sequencing in an independent dataset composed of 307 Caucasian late-onset AD cases and 501 elderly Caucasian controls (p-value= 0.012; OR= 0.54, 95% CI 0.31-0.89). In this follow-up dataset, we confirmed a higher frequency of the ABCA7 p.G215S variant in controls compared to cases (carrier frequency= 13.5% vs 7.8% [1.7-fold], MAF= 7% vs 4.3% [1.6-fold], respectively).

Finally, we report also a common coding polymorphism in ABCA7 (p.R1349Q, rs3745842) in our cohort, that map 1.3 kb from a reported GWAS hit, rs3752246, but cluster within a
different common haplotype block (MAF>5%)(Naj et al., 2011) (Table S2). Rs3745842 major allele (G) was more frequent in cases compared to controls, although the association was nominally significant after multiple testing correction (p-value= 1.4x10\(^{-3}\), corrected p-value=0.081) (Table S6).

### 3.2 Loss of function mutations in \(\textit{ABCA7}\)

Loss of function mutations in \(\textit{ABCA7}\) have been recently reported as main mechanism explaining the GWAS signal and the increased susceptibility to AD.

In our cohort, we detected 5 loss of function mutations in \(\textit{ABCA7}\): 2 stopgain mutations (p.Y1579X and p.E1974X) and 3 splice-site or near splice-site mutations (c.-7-2 A>G, c.-7-7 T>C and c.231-12C>A). \(\textit{ABCA7}\) p.E1974X and c.231-12C>A are novel variants and, together with p.Y1579X, are singletons, detected only in controls (Table S7).

Importantly, the enrichment for \(\textit{ABCA7}\) LoF mutations and novel variants in controls did not rely on the sequencing strategy (exome sequencing vs genome sequencing). These variants have not been indeed mainly detected in the 199 BYU controls, that underwent genome sequencing (Table S8).

Moreover, 3 very rare indels have been identified in controls in the NIH-UCL cohort (p.1402delT, p.1638delCTT and p.1749delCTACTG). \(\textit{ABCA7}\) p.1749delCTACTG is a novel mutation and \(\textit{ABCA7}\) p.1402delT was present also in one case. These indels have been excluded from the pooled dataset because not targeted in the ADNI subcohort (Table S9) (Figure 1).

Finally, 3 nonsense mutations (p.W749X, p.W903X and p.R1754X) and one splice-site mutation, (c.4416+2T>G) did not pass either the sample or variant quality control (QC) criteria and therefore have not been included in the study (Table S10).

### 3.3 Gene-based association test

In addition to single-marker analysis, we carried out gene-wide analysis to combine the joint signal from multiple variants (coding variants and flanking UTRs) within a gene and to
provide greater statistical power than that for single-marker tests. All the variants (nonsynonymous, synonymous, UTRs, singletons) located within the studied genes and their exon-intron flanking regions were collapsed together and their combined effect was studied. \textit{ABCA7} was the main hit both in the SKAT and c-alpha test, nominally and statistically significant, respectively, after multiple testing correction (corrected p-value=0.6 and 5.3x10^{-3} respectively) (\textit{Tables 3a and 3b}). Importantly, given the exclusion of indels in the merged discovery dataset, the presence of only 2 nonsense singleton mutations in \textit{ABCA7} (p.Y1579X and p.E1974X, both detected in controls) and 4 putative splice-site mutations (rs3752229, rs2242437, c.231-12C>A and rs182233998, the latter one nominally significant in controls), our findings are not influenced by a burden of LoF mutations in \textit{ABCA7}. Considering the very rare frequency of these LoF variants, their detection with sufficient power would have required a very large sample size. By contrast, the top signals are represented mainly by common-low frequency coding variants with an higher frequency in controls compared to cases and with a modest to intermediate protective effect (0.329<OR<0.755) (\textit{Table S6}).

4. DISCUSSION

We report the results of single-variant and gene-based association tests performed in \textit{BIN1}, \textit{CLU}, \textit{CR1}, \textit{PICALM}, \textit{MS4A6A}, \textit{ABCA7}, \textit{EPHA1}, \textit{CD33}, \textit{CD2AP} in a cohort composed of 332 apparently sporadic and mainly late-onset AD cases and 676 elderly Caucasian controls from North America and UK. In the single-variant association test we have analysed the effect of low frequency and rare coding variants (MAF<5%), aiming to identify potential functional variant(s) underlying the GWAS hit(s). In the gene-based analysis (SKAT and c-alpha test), we collapsed the full spectrum of variants identified in these loci to study their collective effect. We do not report any pathogenic mutation in \textit{APP}, \textit{PSEN1} and \textit{PSEN2} in our cohort. However, one of the controls was an heterozygous carrier of the protective variant \textit{APP} p.A673T (MAF 7x10^{-4} in our cohort and MAF 5x10^{-4} among the European non-finnish, ExAC database, released 13 January 2015)(Jonsson et al., 2012).
TREM2 p.R47H, the second most common risk factor for sporadic AD, has been detected in 6 cases (1.8%) and 4 controls (0.59%) and, likely given our small sample size, with a MAF=0.2%, was not significantly associated to AD (p-value= 0.09).

ABCA7 was the only significant hit in the c-alpha test and harbors a low-frequency coding variant (p.G215S, rs72973581), whose minor allele confers a modest (OR=0.57, 95% CI= 0.41-0.80) but statistically significant protection (corrected p-value =0.024) against AD. Importantly, this SNP was not present in several GWAS or exome SNP arrays and does not cluster within common haplotypes identified by tagging SNPs, whereas it has been detected through ABCA7 direct sequencing in the current study. Therefore, rs72973581 would have stayed likely undetected using common fine-mapping genotyping arrays. In addition, it does not cluster in the risk haplotypes identified by GWAS main hits (rs3764650, rs115550680, rs3752246, rs4147929), suggesting an independent signal and a likely different pathogenic mechanism of the major allele (Hollingworth et al., 2011)(Reitz et al., 2013)(Naj et al., 2011)(Liu et al., 2014). Importantly, the ABCA7 p.G215S significant protective role against AD is supported by a targeted resequencing study of ABCA7 in a Belgian cohort, where rs72973581 (A) frequency was 1.34-fold higher in controls compared to cases (p-value= 0.055) (Cuyvers et al., 2015). Notably, the main variant associated to LOAD in this Belgian cohort was a low frequency intronic variant (rs78117248) that did not pass our QC filter. However, in line with our findings, Cuyvers et al. report an enrichment for common and low frequency polymorphism with a modest protective role in ABCA7. Importantly, among the top 10 genetic variants identified in our study, 3 missense mutations (rs74176364, rs114782266, rs117187003) have been described associated also to autism spectrum disorder (ASD), strongly pointing towards a functional role of these amino acid changes and suggesting a possible shared pathogenic mechanisms underpinning neurodegenerative and neurodevelopmental diseases (He et al., 2014).

Interestingly, several lines of evidence reported that a significant decrease in ABCA7 levels is associated to AD. At this regard, different and likely not mutually exclusive mechanisms have been described to influence the protein level: 1) common and generally non-coding variants in regulatory regions; 2) alternative splicing; 3) increased CpG island methylation
Recently, also LoF mutations in \textit{ABCA7} have been shown to significantly increase the susceptibility to AD in the Islandic population (Steinberg et al., 2015). This has been replicated in two different populations (Caucasian North American and Belgian) by two independent studies (Vardarajan et al., 2015)(Cuyvers et al., 2015). Therefore we report another potential mechanism, through which low frequency protein coding variability in \textit{ABCA7} may influence AD risk.

Notably, \textit{ABCA7} p.G215S provides critical insights into the genetic architecture of diseases, reinforcing the view that GWAS loci, likewise Mendelian genes, harbor low frequency and rare protective coding variants that can counteract with a similar effect size the damaging alleles (OR \(\approx 0.6\) vs \(\approx 1.1\) and \(\approx 0.2\) vs \(\approx 5\), for GWAS loci and Mendelian genes, respectively) (Nejentsev et al., 2009)(Rivas et al., 2011)(Jonsson et al., 2012)(Asante et al., 2015).

\textit{ABCA7} is mainly expressed in leukocytes and in myelo-lymphatic tissues (thymus, spleen and bone marrow) and microglia in the brain (http://web.stanford.edu/group/barres_lab/brain_rnaseq.html)(http://www.uniprot.org/) (Kim et al., 2008). \textit{ABCA7} encodes for ATP-binding cassette sub-family A member 7 (ABCA7), a multi-pass protein, present on the cell, Golgi and endosome membranes (http://www.uniprot.org/). \textit{In vitro} and \textit{in vivo} experiments have shown ABCA7 pivotal role in phagocytosis and a likely modest role in HDL biogenesis. In \textit{Abca7} \(^-/-\) mice, macrophages and microglia display impaired phagocytosis and clearance of amyloid from the brain, which leads to cognitive impairment (Iwamoto et al., 2006)(Tanaka et al., 2011).

Therefore, \textit{ABCA7}, likewise \textit{TREM2} and \textit{CD33}, may play an important role in regulating microglial uptake and clearance of Aβ debris. \textit{ABCA7} p.G215S clusters within the extracellular topological domain of ABCA7. Remarkably, at the homologous residue, the serine is the reference amino acid in ABCA7 in different mammals and in the homologous protein ABCA1 in humans. Thus, suggesting that this amino acid may confer some biologic advantage and may have been positively selected during the evolution (\textit{Figure S1} and \textit{S2}).
Likewise other low frequency and rare protective variants at the GWAS loci (Table S11), ABCA7 p.G215 is a relatively conserved residue among different species (Figure S1) and this amino acid change (glycine to serine) may only slightly modify the protein activity (-5.86, 56, Gerd and Grantham score, respectively). Moreover, it has been reported as a tolerated change and benign, arguing against any possible loss of function or significant impairment of ABCA7, that have been indeed associated to increased risk for AD (Steinberg et al., 2015). The biological effect of this substitution may therefore lead to a mild ABCA7 gain of function, possibly strengthening the interaction with a binding protein or regulating its expression. Although ABCA7 p.G215 has not been predicted to be a coding target for miRNA (https://www.umm.uni-heidelberg.de), a possible post-transcriptional or post-translational regulation should not be excluded. Importantly, the substitution of a glycine with a serine may imply an additional substrate for serine-kinases or proteases. Moreover, in a similar way, ABCA1, whose LoF variants have been associated to AD (Nordestgaard et al., 2015)(Kim et al., 2012), has been reported to be particularly enriched for low frequency and rare coding variants with an average 1.5-fold higher frequency in controls compared to LOAD cases and a modest protective effect in a Greek cohort (OR =0.96-0.38) (Lupton et al., 2014) (Table S12).

Thus, understanding the effect of ABCA7 p.G215S has the potential of unravelling new pathogenic mechanisms underpinning AD and may provide a promising therapeutic target that would not significantly alter ABCA7 overall physiological function, which is critical for AD development.

Finally, we support Vardarajan et al. resequencing study of the GWAS loci (Vardarajan et al., 2015), confirming a burden of damaging variants in ABCA7 and BIN1 (Table S5) and to a lesser extent in CD2AP, EPHA1 and MS4A6A (main hits in the gene-based analysis) (Table 3a and 3b), highlighting their potential role as susceptibility loci for LOAD.

However, we could not replicate the main hits detected by Vardarajan et al. in the single-variant analysis, either because such variants have been targeted but not detected in our cohort (ABCA7 p.E1679X, EPHA1 p.P460L and BIN1 p.K358R) or because the variants have been targeted but eliminated by the QC filter (CD2AP p.K633R). Thus, suggesting a possible
lack of replication compared to the previous studies attributable to the different population but also different sequencing strategies, capture and coverage. Nevertheless, *EPHA1* and *CD2AP* harbor 2 of the main hits detected in the single-variant analysis in our cohort (rs11768549 and rs143297472, respectively) (*Table 2*), with rs11768549 already associated with the rapid progression of the disease in a cohort of Caucasian North American LOAD cases (Wang et al., 2015).

In summary, we support previous studies, suggesting that 1) *ABCA7* significantly influences AD risk; 2) *ABCA7* p.G215S is likely to reduce the susceptibility to AD; 3) GWAS hits are pleomorphic loci harboring a complex spectrum of variants synergistically contributing to the disease phenotype with different mechanisms, effects (damaging, protective and neutral) and effect sizes (0<OR<4) and 4) gene-based approaches are effective methods to mine genetic data and to accurately filter potential candidate genes.

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We thank contributors, including the Alzheimer’s Disease Centers who collected samples used in this study, as well as patients and their families, whose help and participation made this work possible.

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London Neurodegenerative Diseases Brain Bank and the Manchester Brain Bank from Brains for Dementia Research are jointly funded from ARUK and AS. This work was supported in part by the Intramural Research Program of the National Institute on Aging, National Institutes of Health, Department of Health and Human Services, project number Z01 AG000950-10. Samples from the National Cell Repository for Alzheimer’s Disease (NCRAD), which receives government support under a cooperative agreement grant (U24 AG21886) awarded by the National Institute on Aging (NIA), were used in this study. NIH grant R01 AG042611 to Kauwe J.

CONFLICT OF INTEREST STATEMENT

The authors declare no competing financial or personal interests that can influence the presented work.

REFERENCES


Figure 1. Pipeline of our study design. WES, whole exome sequencing; WGS, whole genome sequencing; INDELS, in frame insertions and deletions; SNVs, single nucleotide variants; CTRLS, controls; UTRs, untranslated regions. * INDELS have been excluded from the discovery cohort because not targeted in the ADNI dataset.
<table>
<thead>
<tr>
<th>COHORTS</th>
<th>N</th>
<th>TYPE</th>
<th>SEQUENCING STRATEGY</th>
<th>ORIGIN</th>
<th>AGE (YRS)</th>
<th>MALE (%)</th>
<th>APOE E4+ (%)</th>
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</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>mean ±SD (range)</td>
<td></td>
<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>(N)</td>
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<td>cases</td>
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<td>Exome sequencing</td>
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<td>Caucasian (North American)</td>
<td>74.68 (60-90)</td>
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<td>Genome sequencing</td>
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<tr>
<td>cases</td>
<td>307</td>
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<td>Sanger sequencing</td>
<td>Caucasian (North American, British, Dutch, Italian, Portuguese)</td>
<td>average &gt;65y</td>
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<td></td>
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<tr>
<td>controls</td>
<td>501</td>
<td>clinical</td>
<td>Sanger sequencing</td>
<td>Caucasian (North American, British, Greek, German, Polish, Australian, Canadian)</td>
<td>&gt;60y</td>
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</table>

Table 1. Description of the different cohorts used in this study. N, number; WES, whole exome sequencing; WGS, whole genome sequencing; YRS, years.
<table>
<thead>
<tr>
<th>GENE</th>
<th>POSITION</th>
<th>MA</th>
<th>cDNA change</th>
<th>Aa change</th>
<th>rs</th>
<th>MAF cases-ctrls (%)</th>
<th>MAF ExAC (%)</th>
<th>SIFT</th>
<th>POLYPHEN</th>
<th>aa/Aa/AA cases</th>
<th>aa/Aa/AA controls</th>
<th>p-value</th>
<th>Corr. p-value</th>
<th>OR (95% CI)</th>
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<tbody>
<tr>
<td>ABCA7</td>
<td>19:1043103 A</td>
<td>c.G643A</td>
<td>p.G215S</td>
<td>rs72973581</td>
<td>4.66-7.24^3</td>
<td>4.31*** Tolerated</td>
<td>Benign</td>
<td>low</td>
<td>0/31/301</td>
<td>1/96/579</td>
<td>0.02</td>
<td>0.8</td>
<td>0.61</td>
<td>(0.38-0.95)</td>
</tr>
<tr>
<td>ABCA7</td>
<td>19:1050996 A</td>
<td>c.G2629A</td>
<td>p.A877T**</td>
<td>rs74176364</td>
<td>0.3-1.18</td>
<td>1.69 Deleterious</td>
<td>Benign</td>
<td>low</td>
<td>0/2/330</td>
<td>0/16/660</td>
<td>0.07</td>
<td>2.8</td>
<td>0.25</td>
<td>(0.02-1.07)</td>
</tr>
<tr>
<td>EPFA1</td>
<td>7:143095153 A</td>
<td>c.G1475A</td>
<td>p.R492Q</td>
<td>rs11768549</td>
<td>2.56-1.47</td>
<td>1.21 tolerated</td>
<td>benign</td>
<td></td>
<td>0/17/315</td>
<td>1/18/657</td>
<td>0.07</td>
<td>2.8</td>
<td>1.86</td>
<td>(0.89-3.84)</td>
</tr>
<tr>
<td>ABCA7</td>
<td>19:1059056 A</td>
<td>c.G5435A</td>
<td>p.R1812H**</td>
<td>rs114782266</td>
<td>1.5-0.81</td>
<td>1.05 Tolerated</td>
<td>Benign</td>
<td>neutral</td>
<td>0/10/322</td>
<td>0/11/665</td>
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<td>6.4</td>
<td>1.87</td>
<td>(0.70-4.92)</td>
</tr>
<tr>
<td>ABCA7</td>
<td>19:1057343 A</td>
<td>c.G4795A</td>
<td>p.V1599M**</td>
<td>rs117187003</td>
<td>0.6-0.22</td>
<td>0.3 Deleterious</td>
<td>Possibly damaging</td>
<td>medium</td>
<td>0/4/328</td>
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<td>2.73</td>
<td>(0.45-18.7)</td>
</tr>
<tr>
<td>CD2AP</td>
<td>6:47573971 A</td>
<td>c.G1488A</td>
<td>p.M496i</td>
<td>rs143297472</td>
<td>0.3-0.07</td>
<td>NA Tolerated</td>
<td>Benign</td>
<td></td>
<td>0/2/330</td>
<td>0/1/675</td>
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<td>10</td>
<td>4.08</td>
<td>(0.21-241.3)</td>
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<tr>
<td>ABCA7</td>
<td>19:1047537 C</td>
<td>c.A2153C</td>
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<td>rs3752239</td>
<td>1.65-2.44</td>
<td>7.02 Deleterious</td>
<td>Benign</td>
<td>low</td>
<td>0/11/321</td>
<td>0/33/641</td>
<td>0.32</td>
<td>12.8</td>
<td>0.66</td>
<td>(0.29-1.37)</td>
</tr>
</tbody>
</table>

**Table2. Most significant variants detected in our discovery set.** Position is in hg19/GRCh37. MAF, minor allele frequency; OR, odds ratio; CI=confidence interval. Corr, Corrected p-value, p-value after Bonferroni correction [p-value*40 [number of variants considered in the single-variant association test]]. *Combined results discovery and follow-up data set. \^MAF cases-ctrls reported a Belgian cohort = 4.66-6.27 (Cuyvers et al., 2015). **Variants reported associated also with autism spectrum disorders (ASD) (He et al., 2014). ***MAF in ExAC (European non Finnish)= 6.14% and MAF in EVS (European American)= 6.24%. \^Variant reported associated to a more rapid disease progression.
<table>
<thead>
<tr>
<th>TRANSCRIPT ID</th>
<th>POSITION</th>
<th>GENE</th>
<th>N.VARIANTS</th>
<th>TEST</th>
<th>P-VALUE</th>
<th>CORRECTED P-VALUE</th>
</tr>
</thead>
<tbody>
<tr>
<td>NM_019112</td>
<td>chr19:1040131..1065563</td>
<td>ABCA7</td>
<td>72</td>
<td>CALPHA</td>
<td>0.000590992</td>
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<tr>
<td>NM_012120</td>
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<td>20</td>
<td>CALPHA</td>
<td>0.0353261</td>
<td>0.31</td>
</tr>
<tr>
<td>NM_152851</td>
<td>chr11:59939123..59950523</td>
<td>MS4A6A</td>
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<td>CALPHA</td>
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<tr>
<td>NM_000573</td>
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<td>CALPHA</td>
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<td>CALPHA</td>
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<td>NM_001831</td>
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<tr>
<td>NM_001772</td>
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<td>CALPHA</td>
<td>0.714286</td>
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*Table 3a. Results from the c-alpha test performed.* Position is in hg19/GRCh37. Corrected p-value, p-value after Bonferroni correction (p-value*9[number of genes considered in the single-gene based analysis]).
<table>
<thead>
<tr>
<th>TRANSCRIPT ID</th>
<th>POSITION</th>
<th>GENE</th>
<th>N. VARIANTS</th>
<th>TEST</th>
<th>P-VALUE</th>
<th>CORRECTED P-VALUE</th>
</tr>
</thead>
<tbody>
<tr>
<td>NM_019112</td>
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**Table 3b. Results from the SKAT test performed.** Position is in hg19/GRCh37. Corrected p-value, p-value after Bonferroni correction (p-value*9[number of genes considered in the single-gene based analysis]).
ABCA7 p.G215S as potential protective factor for Alzheimer’s disease


*The Alzheimer’s Research UK (ARUK) Consortium: Peter Passmore, David Craig, Janet Johnston, Bernadette McGuinness, Stephen Todd, Queen’s University Belfast, UK; Reinhard Heun, Royal Derby Hospital, UK; Heike Kölsh, University of Bonn, Germany; Patrick G. Kehoe, University of Bristol, UK; Emma R.L.C. Vardy, Salford Royal NHS Foundation Trust, UK; Nigel M. Hooper, David M. Mann, Stuart Pickering-Brown, University of Manchester, UK; Kristelle Brown, James Lowe, Kevin Morgan, University of Nottingham, UK; A. David Smith, Gordon Wilcock, Donald Warden, University of Oxford (OPTIMA), UK; Clive Holmes, University of Southampton, UK.

Highlights

• Genome-wide association studies (GWASs) have been effective approaches to investigate common genetic variability underlying complex traits in a systematic and unbiased way.

• Despite the evidence showing the contribution of these loci to AD pathogenesis, their genetic architecture has not been extensively investigated.

• We have used exome and genome sequencing data to analyse the single independent and joint effect of rare and low frequency protein coding variants in 9 AD GWAS loci with the strongest effect sizes after APOE (BIN1, CLU, CR1, PICALM, MS4A6A, ABCA7, EPHA1, CD33, CD2AP) in a cohort of 332 sporadic AD cases and 676 elderly controls of British and North American ancestry.

• We identified coding variability in ABCA7 as contributing to AD risk. This locus harbors a low frequency coding variant (p.G215S, rs72973581, MAF=4.3%) conferring a modest but statistically significant protection against AD (p-value= 6x10⁻⁴, OR= 0.57, 95% CI 0.41-0.80).

• Notably, our results are not driven by an enrichment of loss of function variants in ABCA7, recently reported as main pathogenic factor underlying AD risk at this locus.

• Our study confirms the role of ABCA7 in AD and provide new insights that should address functional studies.