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Article type : Original Article

Title: Genetic signature of human longevity in PKC and NF-κB signaling Short title: PKC and NF-κB signaling in human longevity

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This is the author manuscript accepted for publication and has undergone full peer review but has not been through the copyediting, typesetting, pagination and proofreading process, which may lead to differences between this version and the <u>Version of Record</u>. Please cite this article as <u>doi:</u> 10.1111/ACEL.13362

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ABSTRACT

Gene variants associated with longevity are also associated with protection against cognitive decline, dementia and Alzheimer's disease, suggesting that common physiologic pathways act at the interface of longevity and cognitive function. To test the hypothesis that variants in genes implicated in cognitive function may promote exceptional longevity, we performed a comprehensive 3-stage study to identify functional longevity-associated variants in ~ 700 candidate genes in up to 450 centenarians and 500 controls by target capture sequencing analysis. We found an

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enrichment of longevity-associated genes in the nPKC and NF- κ B signaling pathways by gene-based association analyses. Functional analysis of the top three gene variants (*NFKBIA, CLU, PRKCH*) suggest that non-coding variants modulate the expression of cognate genes, thereby reducing signaling through the nPKC and NF- κ B. This matches genetic studies in multiple model organisms, suggesting that the evolutionary conservation of reduced PKC and NF- κ B signaling pathways in exceptional longevity may include humans.

Keywords: Longevity, Centenarian, Genetic variant, Rare variant, PKC, NF-κB

INTRODUCTION

The heritability of life expectancy is estimated to be ~25% in the general population (Herskind et al., 1996; McGue, Vaupel, Holm, & Harvald, 1993), to lower than 10% considering inherited sociocultural factors (Ruby et al., 2018) in the general population, but it becomes more substantial after age 65 and 85 years, at 36% and 40% (Murabito, Yuan, & Lunetta, 2012), respectively. Family studies suggest that the genetic component of life expectancy is especially strong in the oldest old such as centenarians, who live 100 years or more (Adams, Nolan, Andersen, Perls, & Terry, 2008; Barzilai, Gabriely, Gabriely, Iankowitz, & Sorkin, 2001). These studies support the utility of centenarians as a human model system of exceptional longevity, 'decelerated' aging or 'healthy' agers.

Being a centenarian is rare (only 1 in 5,000 people lives to 100 years in the United States) (Andersen, Sebastiani, Dworkis, Feldman, & Perls, 2012) despite the recent increase in the life expectancy of the general population. In addition to an extended lifespan, centenarians have an extended healthspan via delaying, surviving, or escaping age-associated diseases including cardiovascular diseases, cancer, and neurodegenerative diseases (Ailshire, Beltrán-Sánchez, & Crimmins, 2015; Hitt, Young-Xu, Silver, & Perls, 1999). Thus, exceptional longevity is obviously coupled with exceptional resistance to diseases that lead to earlier mortalities in humans.

Individuals with exceptional longevity manifest delayed onset of Alzheimer's disease (AD) and dementia (Kliegel, Moor, & Rott, 2004; Perls, 2004). Genetic studies indicate that longevity-associated genes may be protective against cognitive decline, dementia and AD (Barzilai, Atzmon, Derby, Bauman, & Lipton, 2006; Christensen, Johnson, & Vaupel, 2006; Dubal et al., 2014; Sanders et al., 2010; Sebastiani et al., 2012). The APOE gene, encoding apolipoprotein E, is a prime example shown to be associated with both AD and longevity (Christensen et al., 2006). Centenarians are depleted of AD-predisposing APOE ɛ4 allele, while they are enriched with the ADprotective APOE ε2 allele. A functional longevity-associated allele in the cholesteryl ester transfer protein (CETP) gene, I405V, is also significantly associated with slower memory decline and lower risk for dementia and AD (Barzilai et al., 2006; Sanders et al., 2010). Longevity-associated KL-VS variants (haplotype including F352V and C370S) in the Klotho gene (KLOTHO) is also associated with protection against cognitive decline (Dubal et al., 2014). A genome-wide association study (GWAS) revealed that variants associated with longevity are most significantly enriched in genes related to AD and dementia (Sebastiani et al., 2012). These results suggest that there are common underlying pathways between longevity and resilience to AD and related cognitive decline. In addition, as shown in the APOE gene, different alleles in the same gene can be associated with opposing outcomes to either accelerate cognitive decline and increase neurodegenerative disease risk or promote longevity and healthy aging in the brain. A rare variant (A673T) discovered in the well-known familial AD causal gene, amyloid precursor protein (APP), protects against AD and related cognitive decline (Jonsson et al., 2012). Taken together, genes associated with cognitive function have emerged as interesting candidate genes that might contribute to human longevity.

In this study, we tested the hypothesis that variants in genes implicated in cognitive function may promote exceptional longevity in humans. We performed a comprehensive 3-stage study to identify functional longevity-associated variants in a total of 701 candidate genes by capture sequencing analysis in up to 450 long-lived individuals (95 years of age or older; defined as centenarians in this study) and 500 controls of Ashkenazi Jewish descent followed by functional studies to ascertain the biological significance of the variant (**Figure 1**). Here we report an enrichment of

longevity-associated genes in the novel PKC (nPKC) and NF- κ B signaling pathways among our candidate genes. Functional analysis of top 3 longevity-associated gene variants (*PRKCH*, *NFKBIA*, *CLU*) *in vitro* suggest that non-coding variants modulate the expression of cognate genes, thereby reducing signaling through the nPKC and NF-kB pathways. Importantly, our hierarchical, multidisciplinary approach led to a genetic signature of human longevity: the tightly-connected and highly conserved PKC and NF- κ B signaling pathways.

Consistent with our data, decreased PKC and NF-κB signaling promotes longevity in model organisms (Curran & Ruvkun, 2007; Spindler, Li, Dhahbi, Yamakawa, & Sauer, 2012; G. Zhang et al., 2013). The directional parallels between the effects of reduction in PKC and NF-κB signaling on longevity of model organisms and those produced by the human longevity-associated *PRKCH*, *NFKBIA*, and *CLU* regulatory variants are quite compelling. These results suggest that reductions in PKC and NF-κB signaling are evolutionary conserved longevity mechanisms in humans and thus represent therapeutic targets for extending healthspan and lifespan.

RESULTS

Discovery of candidate longevity-associated genes and variants by target capture sequencing

GWAS involving long-lived individuals have predicted the presence of rare protective variants with strong effects in centenarians (Deelen et al., 2014; Deelen et al., 2011; Nebel et al., 2011; Sebastiani et al., 2012). Indeed, a comprehensive candidate approach has demonstrated rare functional variants with protective effects are enriched in centenarians (Suh et al., 2008; Tazearslan, Huang, Barzilai, & Suh, 2011), suggesting that centenarians may harbor individually rare, but collectively more common genetic variations in candidate longevity genes. Identifying such rare variants requires extensive sequencing analysis of a large centenarian cohort. To ensure identification of all possible longevity-associated variants, including rare variants, we performed a hierarchical cost-effective sequencing analysis of Ashkenazi Jewish centenarians and controls (**Figure 1**).

In stage 1, we selected 568 candidate genes (**Supplementary Table 1**) implicated in cognitive function. These candidate genes included: 1) genes implicated in neurodegeneration such as AD and Parkinson's disease (PD); 2) AD related genes such as *APP* and *APOE* and interacting pathways; 3) genes implicated in cognitive function such as memory formation mechanism and neuronal receptors; and 4) genes implicated in lipid metabolism such as cholesterol synthesis, transport, and metabolism, e.g. *APOE* and *CETP*.

We then performed target capture sequencing (Capture-seq) of the 568 candidate genes in 51 centenarians and 51 controls. Target regions of candidate genes included exons, exon-intron junctions, and 2kb proximal upstream region to identify both coding and non-coding variants. The Capture-seq analysis identified a total of 13,574 variants in our candidate genes. The variants showed a similar distribution between centenarians and controls across all genomic regions (**Supplementary Table 2**). The vast majority of the variants (74.9%, 10,161 out of 13,574) occurred in non-coding regions with about 20% being previously unreported novel variants.

To identify candidate longevity-associated genes enriched with rare variants, we performed a gene-based association study that considered aggregates of rare variants in a gene region using SKAT analysis (Wu et al., 2011). We identified candidate longevity-associated genes that surpass the threshold of nominal p-value less than 0.05 (**Supplementary Table 3**). We further performed Ingenuity Pathway Analysis (IPA) to identify pathways that were enriched in significant genes (p < 0.05) from the SKAT analysis. We found that almost all of the top enriched pathways (p < 0.05) contained three protein kinase C (PKC) family genes (*PRKCB, PRKCH, PRKCI*) (**Supplementary Table 4**).

Prioritization of candidate longevity-associated genes by genotyping

We performed Fisher's exact test to identify variants associated with longevity. 457 variants were significantly associated with longevity (p < 0.05), among which the most significant variant was rs753381 in *PLCG1* gene (p-value: 0.0003, **Supplementary Table 5**). The majority of variants were rare, accounting for 58.8% (7,984 out of 13,547 variants) with minor allele frequency (MAF) less than 0.05. We then selected variants for genotyping in a larger number of individuals, i.e. 474 centenarians and 551 controls, to prioritize longevity-associated genes, based on the following criteria: 1) association p-values (p < 0.05); 2) rare variants (MAF < 0.05) enriched in either centenarians or controls; and 3) potentially functional variants including non-synonymous, non-sense and frameshift variants with MAF difference between the two groups.

We successfully genotyped a total of 222 variants using Sequenom MassARRAY iPLEX assays (**Experimental procedures**). A total of 23 variants were significantly associated with longevity (nominal p value < 0.05), and 12 variants were enriched in centenarians as compared to controls (**Supplementary Table 6 and 7**). Notably, among the 12 variants, 8 variants were in genes involved in the PKC signaling pathway, including phospholipase C (PLC) family, Ca²⁺/calmodulin-dependent protein kinase (CaMK), and EGF Receptor (*EGFR*). Together with the results from the SKAT analysis and pathway analysis (**Supplementary Table 3 and 4**), this result led us to focus on PKC family genes and genes involved in the pathways that interact with PKC such as PLC and CaMK signaling as top longevity-associated genes among our initial 568 candidate genes implicated in cognitive function.

A second stage Capture-seq analysis focused on PKC and PKC-interacting pathway genes

In stage 2, we selected a comprehensive list of 217 genes (**Supplementary Table 8**) acting in the PKC and PKC-interacting pathways as candidate genes for the second Capture-seq analysis. These included: 1) all known PKC and PLC family genes; 2) genes in PKC-interacting pathways such as PKA, MAPK, and CaMK pathways; 3) genes in PKC-upstream pathways such as neurotrophin receptors, neuronal receptors, EGF receptor, and inflammatory signaling; 4) PKC-downstream genes such as those involved in NF-kB, CREB, NFAT, and mTOR signaling; and 5) AD- and PD-associated genes from GWAS. We also included significant genes (p < 0.05) from the stage 1 Capture-seq study that were not in the PKC interacting pathways.

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We performed the stage 2 second Capture-seq analysis of the 217 genes in 450 centenarians and 500 controls. A total of 23,625 variants were discovered, among which 564 variants were significantly associated with longevity (p < 0.05) and the most significant variant was rs1092331 in *PRKCH* gene (p-value: 0.0001, **Supplementary Table 9**). The vast majority (89.57%, 21,160 out of 23,625) of these variants were in non-coding regions and rare variants (81.05%, MAF < 0.05, **Supplementary Table 10**). Following SKAT analysis, we identified 22 genes that were associated with longevity (nominal p-value < 0.05, **Table 1**).

To determine if any sub-pathways of our PKC pathway-centric candidate genes were enriched with longevity-associated genes, we first sub-categorized all candidate genes and then analyzed the enrichment ratio of longevity-associated genes detected by SKAT analysis in each category (**Figure 2a, 2b** and **Supplementary Table 11**). We found significant enrichment of longevity-associated genes in two sub-pathways (p < 0.05), the novel PKC family (p = 0.024) and the NF- κ B complex (p = 0.024) (**Figure 2a, and 2b**).

In silico analysis to identify potentially functional variants in longevity-associated genes

In Stage 3, we examined the functional relevance of the variants in the longevityassociated genes in the PKC and PKC-interacting pathways. For this purpose, we first conducted *in silico* analysis to prioritize potentially functional variants. For coding variants, we selected variants predicted to affect protein function and/or structure, including non-synonymous variants, stop-gain or stop-loss variants, frameshift variants, and splicing variants. For non-coding variants, we used RegulomeDB (<u>http://regulomedb.org/</u>) to identify potentially regulatory variants. For 3' UTR variants, we took advantage of the starBase database (<u>http://starbase.sysu.edu.cn/</u>), which provides the experimental information of RNA binding protein such as Argonaute (Ago) protein from CLIP-seq analysis. Thus, the genomic location of RNA-binding protein in 3' UTR is used to predict the microRNA binding sites based on experimental results (**Experimental procedures**). Using the predicted functional variants, we performed multiple SKAT analysis of all variants, coding variants, regulatory variants, or combined putative functional variants of the 22 longevity-associated genes (**Table 1**). Along with SKAT analysis, which gives a weight to rare variants and considers both direction of enrichment in two groups, we also performed the SKAT-O and SKAT-C analyses that consider the directionality of rare variants enrichment in either one group and the effects of both common and rare variants, respectively.

Most of the 22 longevity-associated genes detected by SKAT (SKAT-all) also showed significant association with longevity when only predicted functional variants were analyzed by SKAT analysis (**Figure 3a**). Interestingly, most of the significance with functional variants was from the predicted functional effect of the regulatory variants. Notably, in SKAT-C analysis—including both rare and common variants with only coding functional variants—the *PRKCD* gene showed the strongest p-value: 0.000016. The information and the distribution of the *PRKCD* coding variants are indicated in **Supplementary Figure 1**.

We generated protein-protein interaction networks among the 22 longevityassociated genes (**Figure 3b**). The generated networks highlighted PKC (*PRKCD*, *PRKD3*, *PKN2*, *PRKCH*) and their interaction with upstream PLC (*PLCB1*, *PLCG2*), EGF receptor (*EGFR*), and NF- κ B complex that are composed of the transcription factors p65 (*RELA*) and p50 (*NFKB1*) and the repressor I κ B α (*NFKBIA*). In addition, the *CLU* gene that encodes clusterin/ApoJ protein showed overlapping interacting proteins with ApoE protein, alleles of which have been shown to associate with longevity and AD.

We also performed IPA ingenuity canonical pathway enrichment analysis to determine which functional pathways are enriched in longevity-associated genes using a total of 217 candidate genes that we used for Stage 2 Capture-seq analyses (**Supplementary Table 8**). The most significantly enriched pathways are involved in immune response pathways such as pattern recognition receptor in recognition of bacteria and virus (p = 0.0005), the production of nitric oxide and reactive oxygen species in macrophages (p = 0.0027), and the role of macrophages, fibroblasts and endothelial cells in rheumatoid arthritis (p = 0.0039) (**Supplementary Table 12**). The pathways also contain age-related diseases such as signaling of atherosclerosis (p =

0.0045), type 2 diabetes (p = 0.0288), and Huntington's disease (p = 0.0467). Interestingly, we noticed that almost all of the top enriched pathways (p < 0.05) contained PKC family and NF- κ B complex genes (*PRKCD, PRKCH, PRKD3, RELA, NFKB1, NFKBIA*) (**Supplementary Table 12**).

Functional analysis of regulatory variants in longevity-associated genes in cell culture models

Recent studies have shown that the vast majority (>95%) of the variants detected by GWAS are non-coding variants and 65% of non-coding GWAS-associated variants occur in enhancer regions (Hnisz et al., 2013), suggesting that gene regulatory changes contribute to inter-individual differences in genetic risk. Similarly, most of the variants detected by the Stage 2 Capture-seq analyses as well as those in the longevityassociated genes (SKAT p < 0.05) were located in non-coding regions of the genome (**Supplementary Table 10**).

To ascertain biological significance of the longevity-associated gene variants, we performed functional analysis of non-coding variants with predicted regulatory potential using immune cell lines that may be functionally relevant based on the pathway analysis (**Supplementary Table 12**). We selected non-coding variants in the predicted regulatory regions of the two longevity-associated genes *PRKCH* and *NFKBIA* (**Table 2**) in the top sub-pathways (**Figure 2a**), the novel PKCs (*PRKCD, PRKCH, PRKD3*) and NF-κB family members (*NFKBIA, NFKB1, RELA*). We also included non-coding variants in the longevity-associated *CLU* gene (**Table 1 and Table 2**) because of its known function in regulation of PKC pathway through inhibition of the NF-κB pathway (Santilli, Aronow, & Sala, 2003).

Due to extended linkage disequilibrium (LD) in the genome, multiple variants in an associated region can be identified as significant and hence identifying truly causal variants is highly challenging especially for non-coding variants. For example, while an intronic variant, rs1092331, was the most significant variant (p = 0.0001) in the *PRKCH* gene by single variant association analysis (**Table 2**), it has multiple variants in LD ($R^2=1$, **Figure 4a**). We selected 5 candidate genomic regions (E1-E5) harboring longevity-associated variants (V2-V6) or variants (V1, V7, and V8) in high LD that map to predicted enhancers based on well-known histone marks (H3K4me1, H3K27ac) as well as a DNase I hypersensitive site (DHS) in the UCSC genome browser (**Figure 4a**).

To test the functional impact of the putative regulatory variants, we designed and performed enhancer reporter assays using the cloned fragments (E1-E5) harboring wild-type or variant sequences (**Figure 4b**) in the U87-MG human glioblastoma cell line. While E2 and E3 showed enhancer activities (**Supplementary Figure 2a**), only the E2 region harboring V2 (rs2463117, p = 0.0496) and V3 (rs17098729, p = 0.0015) variants showed a modest but significant reduction in reporter activity as compared to E2 harboring the wild-type sequence (**Figure 4c**). Thus, it was expected that centenarian-enriched *PRKCH* gene variants may reduce PKC and NF- κ B signaling through decreased level of *PRCKH* gene expression.

For *CLU* gene (clusterin) variants (**Supplementary Figure 2b**), we selected two candidate regulatory regions (E1 and E2) that map to predicted enhancers based on well-known histone marks (H3K4me1, H3K27ac) as well as DHS in the UCSC genome browser. E1 and E2 harbor the longevity-associated variants (V1: rs9331893 and V2: s9331906, respectively) (**Figure 5a, and Table 2**). We found that these variants are close to the previously reported variant (rs11136000) associated with decreased risk of AD (Harold et al., 2009; Lambert et al., 2009) (**Figure 5a**). Luciferase enhancer reporter assays using reporter constructs (**Figure 5b**) with and without the putative enhancer variants in U87-MG human glioblastoma cell line revealed that the E1 region with the variant rs9331893 had a significant increase in reporter activity compared to the wild-type sequence (**Figure 5c**). Given the inhibitory function of clusterin on the NF- κ B pathway through the stabilization of I κ Bs (Santilli et al., 2003), the increased level of *CLU* expression caused by the longevity-associated variant may diminish PKC and NF- κ B signaling.

The *NFKBIA* gene encoding the NF- κ B repressor I κ B α was the most significantly associated gene by SKAT analysis (p = 0.0053, **Table 1**). We found three centenarianenriched rare variants in the proximal promoter region within 2 kb upstream of transcription start site (**Figure 6a**). The variants were two novel variants, chr14:35875417 (TAGAG>T) and chr14:35874065 (G>A), and rs2233407 (**Table 2**). We found little evidence of LD amongst these variants. To investigate possible changes in the promoter activity conferred by these variants, we generated two promoter reporter constructs (Figure 6b): P1 contained the 1.6kb proximal promoter region that harbors a centenarian-enriched variant, V3 (-1457, chr14:35875417 (TAGAG>T) and P2 contained the 0.6kb proximal promoter region harboring two centenarian-enriched variants, V1 (-105, chr14:35874065 (G>A) and V2 (-563, rs2233407). The P2 region overlaps with a predicted promoter based on the well-known histone mark, H3K4me3, as well as numerous transcription factor bindings (Figure 6a). We performed the *in vitro* reporter assays in THP-1 human monocytic cell line using the P1 and P2 constructs with and without the longevity-associated variants. We found that both chr14:35874065 (G>A) (V1) and rs2233407 (V2) variants in the P1 region caused a modest, but significant increase in promoter activities as compared to wild-type (Figure 6c). Since higher expression of $I\kappa B\alpha$ results in a decrease in NF- κB transcriptional activity, taken together, these results suggest that the longevity-associated variants reduce the signaling of nPKC and NF-kB pathways by modulating expression of their cognate genes.

DISCUSSION

The ultimate goal of this study was to establish if genes implicated in maintaining cognitive function into old age also potentially impact human longevity. From the outset, this study focused on identification of rare, functional variants based on the results from previous GWAS that clearly suggesting the presence of rare, protective variants enriched in long-lived individuals (Deelen et al., 2014; Deelen et al., 2011; Nebel et al., 2011; Sebastiani et al., 2012). Our main hypothesis was that rare functional variants in genes implicated in cognitive function may contribute to human longevity. To test this hypothesis, we conducted a 3-stage systematic study (**Figure 1**). In stage 1 and 2 Capture-seq analysis, we identified 22 longevity-associated genes (**Table 1**), which were found to be enriched in the nPKC family and the NF-κB complex by sub-pathway analysis (**Figure 2a, 2b** and **Supplementary Table11**). By integrating *in silico* analysis and *in vitro* reporter assays in the Stage 3, we identified functional regulatory variants in

three longevity-associated genes, *PRKCH*, *NFKBIA*, and *CLU* that modulated the expression of the cognate genes. The directionality of the effects on gene expression conferred by the longevity-associated variants suggested that these variants decrease signaling through the nPKC and NF-κB pathways, thereby contributing to the longevity phenotype.

Our study represents by far the largest study for a comprehensive genetic analysis of candidate genes, involving 474 centenarians and 551 controls, in search of longevity-associated genes and variants. It should be noted that defining and selecting control individuals pose a unique challenge in any case-control study of human longevity (Sebastiani et al., 2017). This issue becomes more challenging in the identification of rare variants associated with longevity. We have been tackling this critical issue by taking a strategic study design. First, we have been studying Ashkenazi Jewish centenarians and controls, a genetically isolated population which is more powerful in identifying rare causal variants not only for Mendelian disease genes but also complex traits such as Type 2 Diabetes (T2D) (Steinthorsdottir et al., 2014). Second, we have performed functional analyses to confirm the genetic and biological significance of top variants.

In the Stage 1 discovery genetic study, single variant association and genebased association studies with a moderate sample size could detect some genetic signatures at the pathway level. However, it was critical to validate the association in a large number of individuals because only a small portion of significant variants and genes were validated. 6.35% (4 out of 63) of significant variants and 8.7% (2 out of 23) of genes in the Stage 1 Capture-seq remained significant in genotyping and Stage 2 Capture-seq analysis, respectively (**Supplementary Table 7** and **Table 1**). The Stage 2 Capture-seq analysis identified many non-coding region variants, which were mainly intronic (50.96%) and in the upstream region (13.36%), as well as exonic region variants (10.38%). Compared to all variants, a higher proportion of longevity-associated variants were identified in intronic (53.90%) and upstream (16.13%) regions, while lower in the exonic region (9.04%). (**Supplementary Table 10**). This suggests that variants in non-coding regions such as upstream and intronic regions may play an important role in longevity by regulation of their cognate gene expression as was implicated in GWAS studies (Hnisz et al., 2013).

In addition, rare variants (MAF < 0.05) were mainly located in non-coding region including intronic (49.87%), upstream (13.20%), and 3' UTR (17.34%) region (**Supplementary Table 10**). Therefore, we performed gene-based association analysis, SKAT, including the non-coding region, considering their predicted regulatory role of gene expression. With longevity-associated 22 genes (**Table 1**) identified by SKAT analysis including all discovered variants, we also performed multiple SKAT analyses to consider a unique genetic signature of each gene in regards to distribution of variants in genomic region (coding or non-coding), frequency (rare or common), and directionality (one directional or bi-directional) (**Figure 3a**). By this analysis, the overview of unique characteristics of variant in each longevity-associated gene could be predicted. This highlights the importance of setting criteria about which variants need to be included in gene-based association study based on their predicted function.

We defined nominal p values of <0.05 as significant throughout the study in order to include all possible candidate longevity-associated genes and variants to avoid possible type 2 errors. This was particularly important for Stage 1 study with a limited number of centenarians (n=51) and controls (n=51). In stage 2 with a larger number of individuals (500 centenarians and 450 controls), we validated the candidate genes from the stage 1 and could identify the top candidate functional variants. To complement the statistical limitation and boost our confidence, we performed the functional analysis of the PRKCH, NFKB1A, and CLU gene variants. These genes belong to the top candidate pathway, nPKC and NF-kB pathway (Figure 1), which is known to play evolutionary conserved roles in modulating lifespan in model organisms (Curran & Ruvkun, 2007; Spindler et al., 2012; G. Zhang et al., 2013). Indeed, our data indicate that the top candidate variants exert functional effects in regulation of the cognate target gene expression in a direction that is mechanistically parallel to what has been known to promote longevity in the model organisms. Our results suggest that in a well-designed genetic association study of human longevity such as ours, one may not be overly concerned about the inflated type I error rate due to multiple testing as long as (1) the prioritization is not based on p-values only, but also on other genetic and biological

information; and (2) the top variants are confirmed in functional follow-up studies by directly testing for the impact of longevity-associated variants.

Association studies generally leave open the question of whether an associated genetic variant is functionally important or can serve only as a genetic marker with the functional locus co-inherited with the associated variant. To minimize possible errors and spurious associations, we took an approach to assess the functional relevance of gene variations by directly testing their impact in *in vitro* functional analyses (Tazearslan, Cho, & Suh, 2012). For functional studies, we chose physiologically relevant human cell lines for the longevity-associated genes implicated in immune function (**Supplementary** Table 12), the U87-MG human glioblastoma cell line and THP-1 human monocytic cell line, because different cell types have different regulatory mechanisms of gene expression. We chose three longevity-associated genes and their centenarian-enriched variants in the PKC and NF-KB pathways, PRKCH, NFKBIA, and CLU. All but one of the candidate longevity-associated variants in the PRKCH, NFKBIA, and CLU genes were located in non-coding regions of the genome (Table 2). RegulomeDB analysis identified that 4 out of 10 variants were equal or less than 5, which meet the criteria used for the SKAT analysis with functional regulatory variants. In studies using cell model, all of the 4 predicted functional variants affected luciferase reporter activity while the other variants didn't change the activity (Figure 4c, 5c, and 6c), suggesting that functional regulatory variants were predicted well. The longevity-associated CLU variants are close to variants associated with the decreased risk of AD detected by multiple GWAS (Harold et al., 2009; Lambert et al., 2009) (Figure 5a). An AD-associated variant, rs11136000, in the CLU gene was shown to be correlated with reduced incidence of AD as well as increased expression of the CLU1 isoform in the human brain (Ling, Bhongsatiern, Simpson, Fardo, & Estus, 2012). Thus, we hypothesize that the longevity-associated intronic variant contributes to longevity and healthy cognitive aging by increasing CLU gene expression.

In this study, we identified a genetic signature of decreased signaling of PKC and NF-κB pathways that may contribute to human longevity. Interestingly, animal model studies showed that decreased signaling of PKC and NF-kB pathway increased lifespan.

Decrease of the PKC pathway activity (Spindler et al., 2012), novel PKC pathway including PKCn (PRKCH) (Monje, Brokate-Llanos, Perez-Jimenez, Fidalgo, & Munoz, 2011), PKCδ (PRKCD) (Curran & Ruvkun, 2007), and PKD3 (PRKD3) (Feng, Ren, Chen, & Rubin, 2007) have been implicated in longevity in model organisms. Negatively affecting expression of PLC family genes, which are directly upstream of PKC, is also reported to increase lifespan, especially PLCβ (Kawli, Wu, & Tan, 2010) and PLCγ with EGF receptor (Iwasa, Yu, Xue, & Driscoll, 2010). In agreement with the animal studies, PLCB1, PLCG2, and EGFR were found to be associated with longevity in our study. NF-kB complex genes act downstream of PKC. In our study, we identified the 3 major components of NF-κB complex, NFKBIA (IκBα), NFKB1 (p50), and RELA (p65), among the top 22 longevity-associated genes. In model organisms, inhibition of NF-κB activity delayed aging and increased lifespan, while enhanced activity accelerated aging (Kawahara et al., 2009; Tilstra et al., 2012; G. Zhang et al., 2013). Deficiency of CLU gene product in aged mice increases the severity of immune response mediated myocarditis in heart (McLaughlin et al., 2000) and glomerulopathy (Rosenberg et al., 2002), two age-related diseases. Indeed, in *Drosophila*, over-expression of human clusterin (CLU) increased stress resistance and extended lifespan (Lee, Shim, Kang, Park, & Min, 2012). This suggests that this conserved pathway may influence longevity in humans, analogous as to what has been demonstrated in model organisms.

It is possible that the genes and signals we identified are population-specific and may need validations in different human populations (Franceschi, Garagnani, Olivieri, Salvioli, & Giuliani, 2020). However, our data suggest that functional variants with the same directional impact on the evolutionary conserved genes as the variants found in our study may play a similar role in different human populations, providing evidence for therapeutic modulation.

With the identification of rare variants enriched in long-lived individuals, there is now a pressing need to functionally validate the variants to understand the mechanisms by which these variants contribute to longevity in humans. An integrated analysis using experimental and computational approaches in parallel will help elucidate the molecular mechanisms of functional variants, which then be further tested in cell and animal models (Z. D. Zhang et al., 2020). The longevity-associated variants in the context of evolutionary conservation provide unique opportunities to translate genetic discoveries to therapeutic modulation.

In summary, by taking a hierarchical, comprehensive candidate approach, we found a genetic signature of human longevity in the hyperconnected PKC and NF-κB pathways among genes implicated in cognitive function. Our results suggest that reduction in PKC and NF-κB signaling may promote longevity in humans as has been observed in model organisms. Further studies will ultimately reveal the novel role of PKC and NF-κB signaling in longevity and maintenance of cognitive function in human populations and provide important mechanistic insights into the molecular basis of aging.

EXPERIMENTAL PROCEDURES

Study subjects and sample collection

Our study group consisted of 474 Ashkenazi Jewish (AJ) centenarians and 551 AJ controls that were collected as part of a Longevity Genes Project by Dr. Nir Barzilai of the Albert Einstein College of Medicine. A centenarian was defined as a healthy individual living independently at 95 years of age or older and a control was defined as an individual without a family history of unusual longevity. The control group consisted largely of spouses of the offspring of centenarians. All blood samples were rapidly processed to obtain DNA at the General Clinical Research Center of the Albert Einstein College of Medicine. Informed written consent was obtained in accordance with the policy of the Committee on Clinical Investigations of the Albert Einstein College of Medicine.

For the initial discovery by Capture-seq of our candidate genes, we used genomic DNA extracted from immortalized B-lymphocytes; 51 centenarians of 95–109 age range (mean: 99.9 ± 3.8) and 51 controls of 48–81 age range (mean: 66.37 ± 8.79) were used. For Capture-seq and genotyping experiments, whole genome amplified DNA was used for which the template DNA was obtained directly from blood samples and amplified using illustra GenomiPhi V2 DNA Amplification kits (GE healthcare Life Sciences). We studied 450 centenarians of 95–110 age range (mean: 98.21 ± 2.9) and 550 controls of 43–93 age range (mean: 73.14 \pm 9.1) for the Capture-seq and 474 centenarians of 95–110 age range (mean 98.33 \pm 2.92), and 551 controls of 39–94 age range (mean 74.44 \pm 8.77) for the genotyping experiments.

Target selection and generation of customized target capture for the Stage 1 Capture-seq

Genes implicated in cognitive function and related pathway genes were broadly selected as candidate genes for our study. For the initial discovery study by Capture-seq, 568 genes were selected based on a comprehensive literature and database search. For AD and PD genes, we used database AlzGene (<u>http://www.alzgene.org</u>), PDgene (<u>www.pdgene.org</u>), and GWAS catalog (<u>http://www.genome.gov/gwastudies</u>). For the following Stage 2 Capture-seq analyses of prioritized PKC-interacting pathway genes, we selected genes from a comprehensive literature search. The complete list of initial 568 candidate genes for the Stage 1 Capture-seq is provided in **Supplementary table 1** and the lists of the 217 candidate genes for the Stage 2 Capture-seq analysis are in **Supplementary table 8**.

We designed a customized Agilent SureSelect in-solution target capture to enrich our candidate genes. For each selected gene locus, we included a 2-kb upstream region of the transcription start site, all exons, and 20 bp of each exon-intron junction. We used the Agilent eArray (http://earray.chem.agilent.com/earray/) to design and assess coverage across the target genomic regions of bait libraries. We designed an assay of 568 candidate genes for the target capture sequencing experiment to maximize the sequencing capacity (Han et al., 2013). For the 568 genes, design efficiency from eArray was 99%.

Generation of customized target capture for the Stage 2 Capture-seq

For the first Stage 2 Capture-seq, we designed a customized Nimblegen SeqCap EZ Choice library (Roche) for target capture to enrich 30 candidate genes (Supplementary table 8). We included a 2-kb upstream region of the transcription start site, all exons, and exon-intron junction of target genes. We used the NimbleDesign (https://design.nimblegen.com/nimbledesign) with 'Max close match' set to 5 to design and assess coverage across the target genomic regions of bait libraries. For the 30 genes, design efficiency from eArray was 97.9%.

For the second Stage 2 Capture-seq, we designed target capture for 187 candidate genes (**Supplementary table 8**) with NimbleDesign as in the first Stage 2 Capture-seq, except for using 'Preferred close match' set to 3 and 'Maximum close match' set to 17 for update and change of system. Design efficiency from eArray was 96.3%.

Genotyping

The iPLEX MassArray assays (Sequenom) were designed in Genomics Shared Facility at Albert Einstein College of Medicine. A total of 333 SNPs were designed for iPLEX assays in multiplexing groups. 222 SNPs were successfully assayed based on the high quality of peaks visualized in the MassARRAY® Typer software. Using the iPLEX assays, genotyping was performed in 474 centenarians and 551 controls in the Genomics Shared Facility at Albert Einstein College of Medicine. The association analyses were performed using SNP & Variation suite version 7.6.11 (Golden helix) and JMP Genomics (SAS). Statistical differences among groups were assessed by the Fisher's exact test, for both allelotype and genotype distribution comparisons. A two-tailed p-value of < 0.05 was considered significant. Due to the low frequency of the rare variants, we could not consider sex as a covariate in our statistical analysis

Rare variation association analysis

We used a R package called "SKAT" for SKAT, SKAT-O, and SKAT-C analysis as methods for rare variant association from both the Stage 1 and 2 Capture-seq (Wu et al., 2011). The SKAT analysis for the Stage 2 Capture-seq was performed by using minor allele frequency (MAF) of variants in each pool instead of allele counts in each individual using modified R script. The script was generated from personal communication with a person who developed SKAT analysis (Dr. Seunggeun Lee). This method is an alternative way to perform SKAT in Pool-seq result with expense of underestimation of p-value since individual genotype information was lost in a pool.

Selection of predicted functional regulatory variants

As the criteria for selecting predicted functional regulatory variants, we selected variants that have the RegulomeDB score equal to or less than 5, which overlapped with any least one transcription factor binding or DHS. In addition, if at least one experiment reported the Ago protein binding in an overlapping region of variants, we selected the variants predicted to be functional in 3' UTR region.

Cell culture condition

Human glioblastoma cell line U87-MG, and human monocyte cell line THP-1 cells were purchased from American Type Culture Collection (ATCC). U87-MG cells were maintained in Dulbecco's modified Eagle medium (DMEM), THP-1 cells were maintained in RPMI. Both of them were supplemented with 10% fetal bovine serum (FBS) and 1% penicillin-streptomycin at 37°C in a humidified 5% CO₂.

Generation of luciferase reporter constructs and assay

Genomic DNAs from individuals who harbored *PRKCH, CLU* intronic heterozygote variants, and reference sequences were used to amplify the intronic regions by PCR. PCR was performed using Phusion high fidelity DNA polymerase (New England BioLabs, Inc., Massachusetts). The amplified products were cloned into pGL3 vector with SV40 promoter (Promega), and the sequence of the construct was verified. For *NFKBIA* upstream variants, genomic DNA of an individual harboring reference sequence were used for PCR amplification and cloned into promoter-less pGL4 vector (Promega). Mutagenesis was performed to generate constructs with desired variants using QuikChange II XL Site-Directed Mutagenesis Kit (Agilent).

For functional studies of the variants in *PRKCH* or *CLU* genes, 1 x 10⁵ of U87-MG cells were cultured on 24-well plates 24 hours prior to transfection, and transfected with 300 ng consisting of luciferase reporter plasmids and pRL-TK (Promega) in a ratio of 5:1. For functional studies of *NFKBIA* gene variants, 2 x 10⁴ of THP-1 cell were cultured on 96-well plates 24 hours prior to transfection, and transfected with 110 ng consisting of luciferase reporter plasmids and pRL-CMV (Promega) in a ratio of 10:1. Transfections were performed using X-tremeGENE HP transfection reagents (Roche) according to the manufacturer's instructions. 24 hours after transfection, cells were harvested and the luciferase activities were measured using a Dual-Luciferase Reporter Assay System (Promega) on a microplate reader (BioTek). Enhancer and promoter activities were normalized with Renilla luciferase activities.

Protein-protein interaction network generation

We used the Human Protein Reference Database (HPRD, release 9), and Highquality INTeractomes (HINT) as the background network (with all self-interactions ignored) and our selected top genes as the "seeds". To generate the sub-network using Cytoscape program, we searched for all shortest paths between every pair of the "seeds". The resultant shortest paths together constitute the sub-network. To make the sub-network simple and informative, we ignored any shortest paths longer than the smallest shortest path length with which all the selected top genes would be included in the sub-network.

ACKNOWLEDGEMENTS

We would like to thank Genomics Shared Facility at Albert Einstein College of Medicine for their assistance with the iPLEX MassArray assay. All the Illumina sequencings were performed in Axeq Technologies and we would like to thank Dr. Jennifer Bae for the assistance. This work was funded by NIH grants AG024391, AG027734, and AG17242 (Y. S.) and a grant from The Paul F. Glenn Center for the Biology of Human Aging (Y. S.). S. R. is the recipient of a Glenn/AFAR Scholarships for Research in the Biology of Aging. The work was also supported by NIH grant U19 AG056278 to Z. Z., L.J.N., P.D.R., N.B. and Y.S.

CONFLICT OF INTEREST

The authors declare that they have no competing interests.

AUTHOR CONTRIBUTIONS

NJS and YS conceived and designed the experiments. SR and JH performed the experiments. SR, JH, TNK, QZ, SL, ZZ analyzed the data. GA and NB collected and provided AJ samples. SR, LJN, PDR, and YS contributed to the writing of the manuscript. All authors read and approved the final manuscript.

DATA AVAILABILITY

Sequencing data that support the findings of this study have been deposited in NCBI Sequence Read Archive (PRJNA669033, PRJNA669034, and PRJNA669037).

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Supporting Information listing

Supplementary Figure 1. Location and distribution of coding variants in longevityassociated *PRKCD* gene

Supplementary Figure 2. Enhancer activity of the regions used for functional study in the PRKCH gene and distribution of rare CLU gene variants

Supplementary Table 1. The list of 568 candidate genes implicated in impacting cognitive function for the Stage 1 Capture-seq

Supplementary Table 2. Characteristics and numbers of variants in 568 candidate genes by the Stage 1 Capture-seq in 51 centenarians and 51 controls

Supplementary Table 3. Candidate longevity-associated genes from SKAT analysis (p < 0.05) from the Stage 1 Capture-seq, ranked by p value.

Supplementary Table 4. Top Ingenuity canonical pathways enriched with candidate longevity-associated genes detected by SKAT analysis (p < 0.05) in the Stage 1 Capture-seq

Supplementary Table 5. The longevity-associated variants detected by Fisher's exact test (p < 0.05) in the Stage 1 Capture-seq

Supplementary Table 6. The longevity-associated significant variants (p < 0.05) by genotyping of selected variants from the Stage 1 Capture-seq in 474 centenarians and 551 controls

Supplementary table 7. Genotyping analysis result of selected 222 variants from the Stage 1 Capture-seq

Supplementary Table 8. The list of 217 candidate genes for the Stage 2 Capture-seq

Supplementary Table 9. The list of significantly longevity-associated variants (p < 0.05) in 217 candidate genes from the Stage 2 Capture-seq

Supplementary Table 10. Characteristics and numbers of variants in 217 candidate genes from the Stage 2 Capture-seq

Supplementary Table 11. Enrichment analysis to identify sub-pathway genes of PKC and PKC-interacting pathways enriched with longevity-associated genes from the Stage 2 Capture-seq

Supplementary Table 12. Significant IPA canonical pathways (p < 0.05) enriched with significantly longevity-associated genes from the Stage 2 Capture-seq

Supplementary Experimental Procedures

Tables

Table 1 Longevity-associated genes from the Stage 2 Capture-seq in 450 centenariansand 500 controls (p < 0.05)</td>

Gene	SKAT p-value	Gene	SKAT p-value		
NFKBIA	0.0053	NFATC3	0.0294		
CLU	0.0059	PRKAR2A	0.0297		
PRKCD	0.0079	PLCB1	0.0319		
CREB1	0.0097	PDE4B	0.0324		
PRKD3	0.0137	PPP2R1A	0.0327		

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EGFR	0.0216	MAPK4	0.0334		
IL1B	0.0216	MAPK9	0.0391		
PKN2	0.0274	ADCY1	0.0424		
RELA	0.0279	IL6R	0.0457		
NFKB1	0.0287	PRKCH	0.0495		
NFATC2	0.0293	PLCG2	0.0496		



Table 2 Candidate regulatory variants in top prioritized longevity-associated pathwaygenes for *in vitro* functional study

Gene	CHR	POS	ID	Region	MAF- controls	MAF- centen- arians	MAF- 1000G	p-value	Regulome DB Score	LD
PRKCH	Chr14	61995673	rs2463117	Intronic	0.076	0.104		0.0496	4	DB/Seq
PRKCH	Chr14	61996018	rs17098729	Intronic	0.055	0.097		0.0015	7	DB/Seq
PRKCH	Chr14	61997226	rs1088680	exonic-syn	0.078	0.122		0.0033	7	DB/Seq
PRKCH	Chr14	61997393	rs1088679	Intronic	0.065	0.109		0.0013	7	DB/Seq
PRKCH	Chr14	61997531	rs1092331	Intronic	0.063	0.112		0.0004	6	DB/Seq
CLU	Chr8	27467984	rs9331893	Intronic	0.008	0.028		0.0008	4	Seq.
CLU	Chr8	27464081	rs9331906	Intronic	0.008	0.023		0.0085	6	Seq.
NFKBIA	Chr14	35875417	•	Upstream	0.001	0.012		0.0022	6	No
NFKBIA	Chr14	35874523	rs2233407	Upstream	0.023	0.050		0.0018	4	No
NFKBIA	Chr14	35874065	•	Upstream	0.005	0.013		0.0847	3a	No

* Bold numbers indicate RegulomeDB scores less than or equal to 5 that are expected to be functional in our analysis. DB means database reported LD and Seq indicates the LD information from our sequencing result.





Figure 1. Workflow of genetic study to discover longevity-associated genes with functional variants in candidate genes

A 3-stage study design was used for genetic discovery. In Stage 1 with 568 genes identified candidate genes in PKC and interacting pathways for the stage 2. The stage 2 study included 217 genes and identified 22 longevity-associated genes by Captures-seq with larger population. In Stage 3, the top functionally important pathways enriched with longevity-associated genes were selected. The impact of non-coding variants was studied *in silico* and *in vitro* to predict the impact on gene expression and therefore the impact on signaling. Mean with standard deviation of age of each group was indicated.



Figure 2. Sub-pathways of PKC and PKC-interacting pathways enriched with longevity-associated genes from the Stage 2 Capture-seq analyses

(a) 217 sub-pathway genes included in the Stage 2 Capture-seq were categorized and the number of longevity-associated genes (SKAT p < 0.05) (red bar) and non-significant genes (blue bar) in each sub-pathway was shown. Enrichment analysis of longevity-associated genes in each sub-pathway was

performed by Fisher's exact test and -log (p-value) was indicated (black dot) on the graph. Dotted line indicated the threshold of significance of enrichment analysis (p = 0.05). (b) The pathway map indicates the PKC and PKC-interacting genes and sub-pathways (black boxes). Longevity-associated genes (red boxes, SKAT p < 0.05) were indicated below the sub-pathways. Blue circles indicate the top sub-pathways enriched with longevity-associated genes (Figure 2a). Gray boxes indicate sub-pathways without longevity-associated genes.



Figure 3. Gene-based association studies with predicted functional variants and protein-protein interaction networks for the 22 longevity-associated genes

(a) The heatmap represents the significance of each gene-based association analysis that includes SKAT, SKAT-O, and SKAT-C. Each analysis used all variants (-all), predicted functional coding variants (-coding), predicted functional regulatory variants (-regulatory), or combined predicted functional coding and regulatory variants (-combined). The genes were from the Stage 2 Capture-seq analyses found to be longevity-associated (SKAT p < 0.05, Table 1). Black dotted lines divide each SKAT, SKAT-O, and SKAT-C analysis. Gray indicates the absence of variants. (b) The network was generated with the 22 longevity-

associated genes (SKAT p < 0.05) from the Stage 2 Capture-seq analyses. The color of circles indicates the significance from SKAT analysis and the size of circles indicates the significance of SKAT analysis with predicted functional variants. The direct interactions between proteins are indicated by the thick green lines. The red circles represent the top enriched pathways including NF-κB and immune response (left) and PLC, PKC, EGF receptor, and downstream factors (right). Light blue medium-sized circles indicate the proteins included in our candidate gene list without SKAT significance. Green circles indicate the ApoE protein, alleles of which have been associated with longevity and AD in multiple human genetic studies.



Figure 4. Functional analysis of the longevity-associated regulatory variants in the *PRKCH* gene

(a) An overview of *PRKCH* gene region displayed in the UCSC genome browser. Red arrows indicate the longevity-associated variants and green arrows indicate the variants in high LD. E1 to E5 regions indicate the putative enhancer regions harboring the variants to be investigated in our reporter assays. (b) A basic design of enhancer reporter constructs. (c) Enhancer reporter assays using E1 to E5 constructs along with a vector as a baseline control in U87 human glioblastoma cell line. The y-axis indicates relative fold changes in reporter activity of the E1 to E5 constructs harboring either wild-type (WT) or the longevity-associated variants (n=3). * indicates the p-value less than 0.05 by t-test.

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Figure 5. Functional analysis of the longevity-associated regulatory variants in the *CLU* gene

(a) An overview of *CLU* gene region displayed in the UCSC genome browser. Red arrows indicate the longevity-associated variants and blue arrowheads indicate the variants associated with risk of AD from GWAS. E1 and E2 indicate the putative enhancer regions harboring the variants to be investigated in reporter assays. (b) The design of the enhancer reporter constructs. (c) Enhancer reporter assays using

E1, E2 constructs along with a vector as a baseline control in U87 human glioblastoma cell line. The yaxis indicates relative fold changes in reporter activity of the E1 and E2 constructs harboring either wildtype (WT) or the longevity-associated variants (n=3). * indicates the p-value less than 0.05 by t-test.



Figure 6. Functional analysis of the longevity-associated regulatory variants in the *NFKBIA* gene

(a) An overview of *NFKBIA* gene region displayed in the UCSC genome browser. Red arrows indicate the longevity-associated variants. P1 and P2 indicate the putative promoter regions harboring the variants to be investigated in reporter assays. (b) A basic design of promoter reporter constructs. (c) Promoter reporter assays using P1, P2 constructs in THP-1 human monocytic cell line. The y-axis indicates relative fold changes in reporter activity of the constructs harboring either wild-type (WT) or variants (V1, V2, V3) in P1, P2 regions (n=4). * indicates the p-value less than 0.05 by Mann-Whitney U test.

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