# Multi-ethnic transcriptome-wide association study of prostate cancer

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# Abstract

The genetic risk for prostate cancer has been governed by a few rare variants with high penetrance and over 150 commonly occurring variants with lower impact on risk; however, most of these variants have been identified in studies containing exclusively European individuals. People of non-European ancestries make up less than 15% of prostate cancer GWAS subjects. Across the globe, incidence of prostate cancer varies with population due to environmental and genetic factors. The discrepancy between disease incidence and representation in genetics highlights the need for more studies of the genetic risk for prostate cancer across diverse populations. To better understand the genetic risk for prostate cancer across diverse populations, we performed PrediXcan and GWAS in a cohort of 4,769 self-identified African American (2,463 cases and 2,306 controls), 2,199 Japanese American (1,106 cases and 1,093 controls), and 2,147 Latin American (1.081 cases and 1.066 controls) individuals from the Multiethnic Genome-wide Scan of Prostate Cancer. We used prediction models from 46 tissues in GTEx version 8 and five models from monocyte transcriptomes in the Multi-Ethnic Study of Artherosclerosis. Across the three populations, we predicted 19 gene-tissue pairs, including five unique genes, to be significantly (lfsr < 0.05) associated with prostate cancer. One of these genes, NKX3-1, replicated in a larger European cohort. At the SNP level, 110 SNPs met genome-wide significance in the African American cohort while 123 SNPs met significance in the Japanese American cohort. Fine mapping revealed three significant independent loci in the African American cohort and two significant independent loci in the Japanese American cohort. These identified loci confirm findings from previous GWAS of prostate cancer in diverse cohorts while PrediXcan-identified genes suggest potential new directions for prostate cancer research in populations across the globe.

# Introduction

In the past two decades, genome-wide association studies (GWAS) have evolved as a critical method to detect and characterize genomic loci affecting susceptibility to various polygenic disorders. As this important method to detect genetic associations has grown, individuals of non-European ancestries have made up less than 22% of all

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GWAS. Individuals of African, East Asian, and Latin American ancestries made up 2.03%, 8.21%, and 1.13%, respectively [1]. These populations are similarly poorly represented in GWAS of prostate cancer. Individuals of European ancestries make up over 85% of discovery GWAS, while individuals of African, East Asian, and Latin American ancestries make up less than 11%, 3%, and 1%, respectively [2]. Individuals of African American or Latin American ancestries are far more diverse due to more recent genetic admixture compared to individuals with East Asian ancestries [3]. This means that individuals who identify as Latin American or African American may have genomes composed of linkage disequilibrium (LD) blocks from Africa, Europe, and the Americas. Admixture and difference in population structure can lead to complexities in studying genetic associations within diverse populations. Nonetheless, these populations need to be studied to better understand disease risk across the globe. The lack of representation of non-European populations in GWAS can lead to further health disparities from non-transferable findings across populations. Since allele and haplotype frequencies differ across populations, susceptibility to disease will vary with these frequencies [4]. Elucidating this susceptibility has grown increasingly important since many disease risk prediction models lose performance accuracy across populations [5]. Having accurate models to predict disease risk is especially important for prostate cancer since disease risk is noticeably different across populations.

Prostate Cancer is the most commonly diagnosed cancer and second leading cause of cancer death among African American men; one in seven black men will be diagnosed with prostate cancer in his lifetime compared to one in nine white men [6]. Risk factors for prostate cancer include age, family history of disease, and African ancestries. The genetic component of prostate cancer is made of rare variants with high penetrance and many common variants with lower penetrance [7,8]. Adding to the complexity of this analysis, prostate cancer is a remarkably heterogeneous phenotype with various molecular and physical classifications [9]. While prostate cancer susceptibility is increased in African Americans, prostate cancer risk is poorly understood in individuals of East Asian ancestries. The age adjusted incidence rate of prostate cancer in East Asian Americans is nearly three times that of native East Asian individuals [10]. Better understanding of how alleles specific to East Asian populations influence prostate cancer risk could reveal new underlying disease biology. Latin American individuals are the least studied of the three populations. The lack of representation could be due to low rates of incidence of prostate cancer in Central and South America; however, Latin American countries are estimated to have the second highest increase in risk for the disease by 2040 [11].

Little work has been done to understand the underlying genetic effects in prostate cancer in diverse populations. Of the 67 published prostate cancer GWAS in the National Center for Biotechnology Information (NCBI) GWAS Catalog, only 16 studies reported to include individuals of non-European ancestries [12]. One of the largest GWAS published up to this point that included over 140,000 individuals of European ancestries was the Prostate Cancer Association Group to investigate Cancer-Associated Alterations in the Genome (PRACTICAL) Consortium. [8] Two of the largest gene-based association studies specific to prostate cancer were transcriptome-wide association studies (TWAS) of the PRACTICAL GWAS summary statistics [13, 14]. While these studies provided insight into genes associated with prostate cancer in European subjects, they provided little insight into genes affected by ancestry-specific SNPs in diverse populations.

We seek to better understand the genetic architecture of gene expression for prostate cancer in African American, Japanese American, and Latin American populations. To do this, we performed a standard case-control GWAS across 4,769 African American subjects, 2,147 Latin American subjects, and 2,199 Japanese American subjects. In

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addition to GWAS, we performed TWAS using PrediXcan and replicated our data in an S-PrediXcan application to the PRACTICAL summary statistics [15,16]. All scripts and notes for this study can be found at https://github.com/WheelerLab/Prostate-Cancer-Study.

## Methods

#### Genotype and Phenotype Data

Phenotype and genotype data for all individuals in this study were downloaded from the NCBI database of Genotypes and Phenotypes (dbGaP) accession number phs000306.v4.p1. Our project was determined exempt from human subjects federal regulations under exemption number 4 by the Loyola University Chicago Institutional Review Board (project number 2014). Participants were mainly self-reported ethnicities of African Americans, Japanese Americans, or Latin Americans. Cases of cancer within these men were identified by annual linkage to the National Cancer Institute (NCI) Surveillance, Epidemiology, and End Results (SEER) registries in California and Hawaii. Whole genome genotypying was performed on Illumnia Human 1M-Duov3\_B and Human660W-Quad\_v1\_A array platforms surveying 1,185,051 and 592,839 single nucleotide polymorphisms (SNP), respectively [17,18]. The final association analyses included 4,769 African American subjects (2,463 cases and 2,306 controls), 2,147 Japanese American subjects (1106 cases and 1093 controls), and 2,199 Latin American subjects (1081 cases and 1066 controls) (Table 1).

**Table 1. Cohort characteristics:** Genotype and phenotype data from the three populations went through standard genome-wide quality control and genotype imputation.

Population	African American	Japanese American	Latin American
Pre-QC Individuals	4874	2199	2147
Post-QC Individuals	4769	2199	2147
Cases	2463	1106	1081
Controls	2306	1093	1066
Pre-QC SNPs	1,199,187	657,366	657,366
Post-QC SNPs	1,077,583	540,326	539,366
Post-Imputation SNPs	15,394,464	4,623,264	7,010,834

#### Quality Control and Imputation

After we downloaded the data from dbGaP, we divided the subjects into three groups of their self-reported ethnicities. Standard genome-wide quality control was performed separately on each of the three groups using PLINK [19]. In each group, we removed SNPs with genotyping rates < 99%. We then removed SNPs significantly outside of Hardy-Weinberg Equilibrium ( $P < 1 \times 10^{-6}$ ). We also filtered out individuals with excess heterozygosity, removing individuals at least three standard deviations from mean heterozygosity. We then used PLINK to calculate the first ten principal components of each cohort when merged with three populations from HapMap phase 3 [20]. The first three principal components of each group were used to confirm self-identified ethnicity (S1 Fig).

Following the confirmation of ethnicity, filtered genotypes were imputed using the University of Michigan Imputation Server [21]. All three groups of genotypes were imputed separately using minimac4 and Eagle version 2.3 for phasing. For the Japanese

American cohort, 1000 Genomes Phase 3 version 5 (1000G) was used as a reference panel [22]. Both the African American and Latin American groups were imputed with 1000G and the Consortium on Asthma among African-ancestry Populations in the Americas (CAAPA) [23]. Genotypes for all three cohorts were filtered for imputation accuracy and minor allele frequency (MAF). SNPs with  $r^2 < 0.8$  and MAF < 0.01 were removed from the analysis. The union of genotypes imputed with 1000G and CAAPA in African American and Latin American cohorts were used for downstream analysis.  $R^2$ and MAF filtering took place before the merging of genotypes. For SNPs in the intersection of the two imputation panels, the SNP imputed with CAAPA was selected for the GWAS and PrediXcan analysis due to LD similarity between the CAAPA reference panel and the populations studied. The number of SNPs and individuals in each ethnicity at various steps throughout quality control and imputation are reported in Table 1.

#### Genome-Wide Association Study

We performed a traditional case-control GWAS of prostate cancer using a logistic regression in PLINK. The first ten principal components were used as covariates to account for global population structure.  $P < 5 \times 10^{-8}$  was used as the P-value threshold to denote genome-wide significance. Independent loci were determined and analyzed using deterministic approximation of posteriors for GWAS (DAP-G), an integrative joint enrichment analysis of multiple causal variants [24]. SNPs were clustered into groups with both a cluster posterior inclusion probability (PIP) and an individual SNP PIP. SNPs in the same cluster were identified as linked, and each group of SNPs was considered a locus independent of others. Pairwise LD calculations of  $r^2$  were calculated in PLINK [19].

#### PrediXcan and S-PrediXcan

We used the gene expression imputation method, PrediXcan, to predict genetically regulated levels of expression for each individual across the three cohorts [15]. We predicted expression using five models built from monocyte transcriptomes of self-identified European (CAU), Hispanic (HIS), African American (AFA) individuals in the Multi-Ethnic Study of Atherosclerosis (MESA). The two other MESA models were built from combined African American-Hispanic (AFHI) data, and all three populations (ALL) [25]. Additionally, we also predicted expression using the 46 multivariate adaptive shrinkage (mashr) models built from 46 tissues in the Gene-Tissue Expression Project (GTEx) version 8 [26,27]. Ovary, uterus, and vagina were excluded from the total tissues. In the GTEx version 8 prediction models, only tissues from individuals with European ancestries were used. All gene expression prediction models may be found at http://predictdb.org/. We tested the predicted expression levels for association with the case-control status of individuals in each ethnic cohort using the first ten principal components as covariates. Significant gene-tissue association were determined after performing an adaptive shrinkage using the R package ashr to account for multiple testing [28]. The adaptive shrinkage calculated a local false sign rate (lfsr)for each test. Gene-tissue pairs with lfsr < 0.05 were considered significant. We chose *lfsr* over traditional false discovery rate because *lfsr* accounts for both effect size and standard error for each gene-tissue pair. To confirm significant gene-tissue findings, we used COLOC to investigate colocalization between the GWAS and expression quantitative trait loci (eQTLs) [29]. We followed up these finding using the summary statistics version of PrediXcan, S-PrediXcan [16]. We applied the same GTEx prediction models to the PRACTICAL summary statistics, which included over 20.3M SNPs [8].

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### Results

#### Genome-Wide Association Studies

To better characterize the genetic architecture underlying prostate-cancer across non-European populations, we performed a case-control GWAS of prostate cancer across 15M SNPs in nearly 5,000 self-identified African American individuals. Additionally, we performed GWAS across 4.6M SNPs in a nearly 2,200 Japanese American individuals and 7.0M SNPs in 2,147 Latin American individuals. Notably, this GWAS includes imputed SNPs from 1000G and CAAPA, two reference panels not available at the time of the original studies of this cohort [17, 18, 22, 30].

In our GWAS of 4,769 self-identified African American individuals, we found 110 SNPs to be significantly associated ( $P < 5 \times 10^{-8}$ ) with prostate cancer. Of these 110 SNPs, 108 of them were located at a previously identified region of chromosome 8 (Fig 1A). Of the SNPs on chromosome 8, rs7659456 was the most significantly associated at  $P = 1.01 \times 10^{-15}$ . The minor allele (T) of rs7659456 is a SNP found only in individuals of recent African ancestries (Fig 2). Four independent clusters were identified by DAP-G on chromosome 8 at 8q24 (Fig 1C). Of the four clusters, two contained SNPs meeting genome-wide significance, one led by rs7659456 (PIP=0.942) and the other led by rs72725879 (PIP=0.994) (S1 Table). rs72725879 was also significant in the Japanese American GWAS (Fig 1B,D). Two of 110 SNPs (rs10149068 & rs8017723) were found at a novel locus on chromosome 14. These two SNPs on chromosome 14 are linked ( $r^2 = 0.989$ ) and identified as one locus.

The GWAS of 2,199 Japanese American individuals identified 123 SNPs as 185 genome-wide significant. Every genome-wide significant SNP was located at the 8q24 186 region of chromosome 8 (Fig 1B). rs1456315 was the most significantly associated SNP 187 in this study  $(P = 1.40 \times 10^{-13})$ . We identified six distinct clusters of SNPs with 188 DAP-G (Fig 1D). Two of the six clusters contained SNPs meeting genome-wide 189 significance with PIP > 0.5. rs1456315 had not only the lowest P-value, but also the 190 highest PIP (PIP = 0.990) in the Japanese American GWAS (Fig 2). rs1456315 was 191 found to be marginally significant in the African American GWAS ( $P = 1.29 \times 10^{-7}$ ). 192 rs1456315 and rs72725879 are linked  $(r^2 = 0.815)$  in the Japanese American GWAS 193 cohort and have similar allele frequencies across East Asian populations while rs1456315 194 and rs72725879 are less linked  $(r^2 = 0.448)$  in the African American GWAS cohort and 195 have divergent allele frequencies across African populations (Fig 2). The GWAS of 7M 196 SNPs in 2,147 Latin American individuals identified no genome-wide significant SNPs. 197 chr13:106685795 was the most associated SNP ( $P = 4.41 \times 10^{-7}$ ), located on 198 chromosome 13 (S2 Fig). 199

Fig 1. Fine-mapping of the top prostate cancer GWAS signals in African Americans and Japanese Americans. (A & B) depict a LocusZoom plots of GWAS results from African American and Japanese American populations, respectively [31]. The most significant SNPs in both GWAS were in the same chromosome 8 region. (A) is plotted using 1000G AFR 2014 LD, and (B) is plotted using 1000G ASN 2014 LD. The y-axis is the  $-log_{10}(P)$  while the x-axis is location on chromosome 8 measured in megabases (Mb). Color represents the LD  $r^2$ . (C & D) depict the GWAS  $-log_{10}(P)$  compared to DAP-G SNP posterior inclusion probabilities (PIP) for the African American and Japanese American populations, respectively [24]. Each point on the plot represents one SNP in each GWAS. The color of each point represents the independent cluster the SNP was assigned to in its respective population. Grey points represent those that were not clustered by DAP-G. 164

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Fig 2. Global allele frequencies of SNPS significantly associates with prostate cancer. A depiction of the global minor allele frequencies rs7659456 (A), rs72725879 (B), and rs1456315 (C). (A) rs7659456 represents the most significantly associated SNP in the African American GWAS. The minor allele, T, is found only in populations of recent African ancestries. (B) rs72725879 represents the only SNP to be identified by DAP-G in a cluster in both the African American and Japanese American GWAS. (C) rs1456315 represents the most significantly associated SNP in the Japanese American GWAS. rs1456315 (C) is found in strong LD with rs72725879 (B) ( $r^2 = 0.815$ ) in the Japanese American GWAS cohort. rs1456315 and rs72725879 are not linked in the African American GWAS cohort ( $r^2 = 0.448$ ). This figure was adapted from one generated using the Geography of Genetic Variants Browser [32].

#### Gene-Based Association Studies

We used the gene expression imputation tool, PrediXcan to predict gene expression from genotypes using models built from transcriptomes of 46 tissues in version 8 of GTEx [26,33]. We also predicted gene expression using five models built from monocyte transcriptome of diverse populations in MESA [25]. After predicting the genetically regulated level of expression for each gene using each of these models, we tested the predicted gene expression for association with the phenotype status of each subject using the first ten principal components as covariates.

Fig 3. Prostate Cancer PrediXcan Results for GTEx predicted genes in African American and Japanese American Populations. (A & B) are Manhattan plots of the PrediXcan results using GTEx version 8 mashr gene expression prediction models for the respective African American and Japanese American populations. Each point represents a gene-tissue test for association with prostate cancer via PrediXcan. The y-axis represents the  $-log_{10}(P)$  of the gene-tissue test, and the x-axis plots chromosome number. The size of the dot is inversely proportional to its *lfsr*.

In our application of PrediXcan to the 4,769 African American individuals, we 208 predicted expression of 489,459 gene-tissue pairs. We identified two gene-tissue pairs 209 with lfsr < 0.05 and nine gene-tissue pairs with lfsr < 0.10 across all GTEx version 8 210 mashr prediction models (Table 2; Fig 3). EBPL was the gene in all nine of the 211 gene-tissue pairs. The two most significantly associated gene-tissue pairs by *lfsr* were 212 found in cerebellar hemisphere (lfsr = 0.0423) and cervical spinal cord (lfsr = 0.0485). 213 The gene-tissue pair with the lowest P-value was KLK3, a gene encoding for 214 prostate-specific antigen, in Tibial Artery  $(P = 3.31 \times 10^{-6})$ , but the lfsr was not 215 significant (lfsr = 0.921). No gene-tissue pairs in the MESA prediction models 216 significantly associated with prostate cancer in the African American population (S3 217 Fig). 218

We used PrediXcan to impute and associate gene expression with prostate cancer across 270,813 gene-tissue pairs in GTEx version 8 from 2,199 Japanese American individuals. We found seventeen gene-tissue pairs to be significantly associated with prostate cancer in this population (Table 2). Of these seventeen gene-tissue pairs, four unique genes were identified: *PLCL1*, *NKX3-1*, *FAM227A*, *COQ10B*. The most significantly associated gene-tissue pair by P-value was *COQ10B* in Thyroid  $(P = 2.28 \times 10^{-7})$ . When we applied PrediXcan to the Latin American cohort, we predicted expression of 411,366 gene-tissue pairs in GTEx version 8. No genes were significantly associated with prostate cancer by *lfsr* or P-value (S4 Fig).

We attempted to replicate our PrediXcan findings by applying S-PrediXcan to GWAS summary statistics of the PRACTICAL meta-analysis of over 140,000

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PRACTICAL summar	y statistics. I	Beta(PRACTICAL) represents the effect dir	rection an	id size pred	icted fron	a S-PrediXcan. "NA" n	neans that the p
gene was not tested in	the <b>PRACT</b>	ICAL summary statistics.					aC
Population	Gene	Tissue	$_{ m lfsr}$	Р	$\mathbf{Beta}$	P (PRACTICAL)	Beta (PRACTIC
African American	EBPL	Brain_Cerebellar_Hemisphere	0.0423	5.25 E-05	-0.022	0.3	Y 900.0
African American	EBPL	Brain_Spinal_cord_cervical_c-1	0.0485	4.94E-05	-0.032	0.551	0.005
Japanese American	PLCL1	Adrenal_Gland	0.0448	9.72 E-07	0.752	0.419	890.0-
Japanese American	PLCL1	Artery_Aorta	0.0435	9.72 E-07	0.738	0.419	nati
Japanese American	NKX3-1	Brain_Caudate_basal_ganglia	0.0472	1.05 E-05	-0.208	3.61 E- 25	-0.288
Japanese American	FAM227A	Brain_Nucleus_accumbens_basal_ganglia	0.042	9.13E-06	-0.211	NA	l lic
Japanese American	NKX3-1	Brain_Putamen_basal_ganglia	0.0475	1.05 E-05	-0.212	1.70E-46	-0.215
Japanese American	FAM227A	Brain-Putamen-basal-ganglia	0.0379	8.21E-06	-0.183	NA	e. VN
Japanese American	NKX3-1	Brain_Substantia_nigra	0.0209	$3.80 \text{E}{-}06$	-0.238	1.18E-20	-0.288
Japanese American	FAM227A	Esophagus_Mucosa	0.0444	9.80E-06	-0.206	NA	NA
Japanese American	NKX3-1	Heart_Left_Ventricle	0.0469	1.05 E-05	-0.202	$3.61 \text{E}{-}25$	-0.28
Japanese American	FAM227A	Heart_Left_Ventricle	0.0434	9.22 E-06	-0.222	NA	NA
Japanese American	NKX3-1	Liver	0.0468	1.05 E-05	-0.198	$3.61 \pm 25$	-0.274
Japanese American	PLCL1	Lung	0.0485	9.72 E-07	0.797	0.419	-0.072
Japanese American	NKX3-1	Muscle_Skeletal	0.0413	8.02 E-06	-0.147	$3.61 \pm 25$	-0.375
Japanese American	PLCL1	Nerve_Tibial	0.031	1.40E-06	0.44	0.537	-0.033
Japanese American	FAM227A	Pituitary	0.0418	9.09 E - 06	-0.178	NA	NA NA
Japanese American	FAM227A	Prostate	0.045	1.00E-05	-0.186	NA	NA
Japanese American	COQ10B	Thyroid	0.0418	2.29 E-07	-1.138	0.298	0.13

able 2. PrediXca	n Significant	<b>Genes</b> Significant $(lfsr < 0.05)$ gene-tis	issue pairs	from PrediXc <sup>i</sup>	an analysis.	<i>lfsr</i> represents the	local false sigr
te as calculated usi	ng adaptive sh	urinkage [28]. P(PRACTICAL) represent:	ts P-value	for the gene-ti	issue pair in	the S-PrediXcan a	unalysis of the
RACTICAL summa	ry statistics. I	3eta(PRACTICAL) represents the effect of	direction a	und size predic	ted from S-	PrediXcan. "NA" 1	neans that the
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Table 2. PrediXcan	Significant	: Genes Significant $(lfsr < 0.05)$ gene-tissu	ue pairs fi	rom PrediXc	an analysis	s. <i>lfsr</i> represents	s the local fal	š
rate as calculated usin	g adaptive sl	nrinkage [28]. P(PRACTICAL) represents F	P-value fo	or the gene-t	issue pair i	in the S-PrediX	can analysis	Ö
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individuals of European ancestries performed by Schumacher et al. [8]. 424,518 230 gene-tissue pairs were tested from the summary statistics using GTEx version 8 231 prediction models. When we compared these summary level results to our findings, only 232 NKX3-1 replicated with a P-value meeting Bonferroni significance ( $P < 1.178 \times 10^{-7}$ ). 233 Similar to the findings of our TWAS, the direction of effect predicted by S-PrediXcan 234 was negative, associating decreased expression of NKX3-1 with prostate cancer (Table 235 2). Not enough SNPs were not present in the FAM227A prediction model to reliably 236 predict expression from PRACTICAL summary statistics. 237

## Discussion

Broadly, the findings of this study confirm previously well-established information about 239 the genetics of prostate cancer in diverse populations. Two of these principal findings 240 were identifying risk loci at chromosome 8q24 and the identifying NKX3-1 as a risk 241 gene for prostate cancer. Prostate cancer risk at 8q24 has been well characterized to 242 carry SNPs that are population-specific to African Americans [18,34]. Previously, up to 243 twelve independent risk signals have been identified at 8q24 in European 244 populations [35] Our study found four independent clusters of SNPs at this position for 245 the African American cohort and six clusters at the Japanese American cohort using 246 DAP-G. The small number of independent signals identified by DAP-G is unsurprising 247 since DAP-G is a more conservative finemapping tool that assumes a single causal 248 variant is expected a priori. Interestingly, of the 102 SNPs assigned to clusters by 249 DAP-G in either African or Japanese American population, only rs72725879 overlapped 250 across populations (Fig 2). rs72725879 has previously been implicated in Asian 251 ancestry-specific risk to prostate cancer [36], and it is found in high LD ( $r^2 = 0.815$ ) 252 with rs1456315, the most significantly associated SNP in our Japanese American GWAS. 253 Additionally, it has been associated with prostate cancer in African Americans [37]. 254

#### rs76595456 is identified as an African Ancestry-specific SNP

rs76595456 was the most significantly associated  $(P = 1.01 \times 10^{-15})$  SNP in our study. Its minor risk allele is specific to populations of recent African ancestries (MAF = 0.1150), and it is absent in both Asian and European populations [22] (Fig 2). The SNP had an individual PIP of 0.942 and was found in a cluster with seven other SNPs bearing a cluster PIP of 0.995. The SNP is an intron variant of *PCAT2*, a well-established prostate cancer associated transcript [37].

#### Novel gene associations implicated by TWAS

In our TWAS of prostate cancer across African Americans, we report EBPL as a gene significantly associated (lfsr = 0.0423) with prostate cancer in this population. EBPLmade up the top twenty gene-tissue pairs by lfsr (lfsr ranging from 0.0423-0.153). It was the most associated gene reported across all five MESA gene expression prediction populations by both P-value (P = 4.67e - 6 in AFA) and lfsr (lfsr = .106 in AFHI). The highest gene-tissue COLOC P4 value, the probability that the GWAS and eQTL

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signal are colocalized, was 0.18. In the S-PrediXcan analysis of the PRACTICAL data, the gene did not replicate. This failure to replicate could be attributed to difference genetic architecture across the African American test population and the European PRACTICAL population.

The TWAS of prostate cancer in the Japanese American population revealed four unique genes associated with prostate cancer. The previously discussed NKX3-1 is a well-established tumor suppressor gene whose decreased expression is known to progress the aggressiveness of the tumor [39, 41]. COQ10B has not been associated with prostate cancer previously according to the NCBI GWAS Catalog [12]. PLCL1 has been nominally associated with prostate cancer through a SNP x SNP interaction study [42]. Neither of these genes replicated in the larger S-PrediXcan analysis. FAM227A provides an interesting situation since it has been previously identified to be associated with prostate cancer in a GWAS of prostate cancer in Middle Eastern populations [43]. Additionally, it was the only significant gene-tissue pair to be found in prostate tissue. SNPs were not present in the PRACTICAL summary stats to generate a reliably predicted level of expression of FAM227A using S-PrediXcan. Despite having nearly 15 million more SNPs in the PRACTICAL summary statistics compared to our GWAS and TWAS of Japanese American populations, SNPs were not genotyped or imputed at this location on chromosome 22 in the PRACTICAL study. One of the SNPs in this model, rs16999186, has a divergent allele frequency across Japanese (MAF = 0.114) and European (MAF=0.0169) populations [22]. Since this frequency in European populations falls near the quality control MAF threshold in the original PRACTICAL study and below the genotyping threshold for the European-specific designed genotyping array, this SNP could easily be missed in larger studies [8]. FAM227A also lies within approximately 150kB of SUN2 ( $P = 1.18 \times 10^{-5}$ ; lfsr = .108) a gene marginally significant in our MESA-HIS prediction model. SUN2 has recently been identified as a gene whose decreased expression has been significantly associated with prostate cancer in Japanese populations [44].

# Conclusion

In summary, this study of 4,769 African American and 2,199 Japanese American individuals identifies potential population-specific risk loci for prostate cancer in people of recent African or East Asian ancestries. Since its minor risk allele is found only in populations of recent African ancestries, rs7659456 is suggested as a potentially novel risk SNP to prostate cancer in African Americans. Furthermore, the identification of FAM227A as a potential susceptibility gene for prostate cancer in non-European populations highlights growing need for studies of the genetics of prostate cancer in non-European populations. This study's principal limitations were sample size and ancestry matching in gene expression prediction models. Sample sizes of under 5,000 African American subjects and nearly 2,200 Japanese American subjects pale in comparison to studies of exclusively European populations that are nearly two orders of magnitude larger. Considering that African American men are nearly twice as likely to die from prostate cancer as their white counterparts, the need for larger sample sizes of African American subjects cannot be overstated [6]. Regarding the gene expression prediction models used, the GTEx version 8 prediction models are the most comprehensive set of prediction models to date; however, they are built exclusively from the transcriptomes of European ancestries individuals. Where the models provide accuracy and breadth in capturing common eQTLs, they struggle to predict expression from population-specific eQTLs. The MESA prediction models used capture some of the diversity across populations, but they too are limited by sample size (233 African American individuals, 352 Hispanic individuals, and 578 European individuals) and

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diversity considering there is no model built from transcriptomes of East Asian subjects [25]. To better understand the genetic processes that underlie prostate cancer in diverse populations, more diverse studies are needed.

# Supporting information

S1 Fig. Quality Control PCA against HapMap. After merging genotypes with 330 those of four reference populations from version three of the HapMap Project, we 331 performed principal component analysis of all three study populations separately. 332 African American (A) and Japanese American (B) genotypes are plotted with three 333 populations from HapMap: Chinese in Beijing and Japanese in Tokyo (ASN), European 334 ancestries in Utah (CEU), and Yoruba people in Ibadan, Nigeria (YRI). The Latin 335 American genotypes are plotted with Chinese in Beijing and Japanese in Tokyo (ASN), 336 European ancestries in Utah (CEU), and indigenous people of North America (NAT). 337

**DAP-G Clustered SNPs** At chromosome 8q24, DAP-G calculated S1 Table. 338 PIPs for 12,785 SNPs and 3,878 SNPs in the African American and Japanese American 339 populations, respectively. Of these SNPs with a calculated PIP, 223 SNPs (24 SNPs in 340 African Americans and 199 SNPs in Japanese Americans) were placed into a cluster. Of 341 those placed into a cluster, 102 SNPs at chromosome 8q24 met genome-wide significance 342 in either population. Nine SNPs in the African American cohort and 93 SNPs in the 343 Japanese American cohort met genome-wide significance and were placed into a cluster. 344 Of these 102 clustered SNPs meeting genome-wide significance, rs72725879 was the only 345 SNP to overlap across populations. This table contains DAP-G results for the 102 SNPs 346 clustered that met genome-wide significance. 347

S2 Fig. Chromosome 13 GWAS & DAP-G Results. Genome-wide association 348 studies identified no genome-wide significant SNPs. (A) depicts a LocusZoom plots of 349 the most associated GWAS results from Native American population on chromosome 350 13 [31]. (A) is plotted using 1000G AMR 2014 LD. The y-axis is the -log(P) while the 351 x-axis is location on chromosome 13 measured in megabases. Color represents the LD 352  $r^2$ . (B) depicts the results of our GWAS in comparison to DAP-G cluster and PIP for 353 the Latin American population [24]. Each point on the plot represents one SNP in our 354 GWAS. The y-axis is  $-\log(P)$ , and the x-axis is the individual SNP PIP as calculated by 355 DAP-G. The color of each point represents the cluster to which DAP-G assigned it. 356

**S3 Fig. MESA PrediXcan Manhattan Plots.** (A, B, & C) are Manhattan plots of the gene-based association study using MESA monocyte gene expression prediction models for the respective African American, Japanese American, and Latin American populations. Each point represents a gene-tissue test from PrediXcan. The y-axis represents the -log(P) of the gene-tissue test, and the x-axis plots chromosome number. The size of the dot is inversely proportional to its *lfsr.* 360

**S4 Fig.** Latin American PrediXcan Manhattan Plots.Manhattan plot of the gene-based association study using GTEx version 8 mashr gene expression prediction models for the Latin American cohort. Each point represents a gene-tissue test from PrediXcan. The y-axis represents the -log(P) of the gene-tissue test, and the x-axis plots chromosome number. The size of the dot is inversely proportional to its *lfsr*.

# Acknowledgments

This work is supported by the NIH National Human Genome Research Institute Academic Research Enhancement Award R15 HG009569 (HEW), the Loyola Carbon Undergraduate Research Fellowship (PNF), the Loyola Mulcahy Scholarship (PNF, JDM, MAS), and the Loyola Provost Fellowship (JDM). Funding support for the GENEVA Prostate Cancer study was provided through the National Cancer Institute (R37CA54281, R01CA6364, P01CA33619, U01CA136792, and U01CA98758) and the National Human Genome Research Institute (U01HG004726). Assistance with phenotype harmonization, SNP selection, data cleaning, meta-analyses, data management and dissemination, and general study coordination, was provided by the GENEVA Coordinating Center (U01HG004789-01). The content is solely the responsibility of the authors and does not necessarily represent the official views of the National Institutes of Health.

The datasets used for the analyses described in this manuscript were obtained from dbGaP at http://www.ncbi.nlm.nih.gov/projects/gap/cgi-bin/study.cgi? study\_id=phs000306.v3.p1.

# Author Contributions

Peter N. Fiorica conceived and designed the experiments, performed the experiments, analyzed the data, contributed reagents/materials/analysis tools, prepared figures and/or tables, authored or reviewed drafts of the paper, approved the final draft.

Ryan Schubert performed the experiments, reviewed drafts of the paper, approved the final draft.

John D. Morris and Mohammed Abdul Sami performed the experiments, contributed reagents/materials/analysis tools, approved the final draft.

Heather E. Wheeler conceived and designed the experiments, analyzed the data, contributed reagents/materials/analysis tools, authored or reviewed drafts of the paper, approved the final draft.

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Fig 1





Fig 2





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