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## Data article

# Data on genetic associations of carotid atherosclerosis markers in Mexican American and European American rheumatoid arthritis subjects

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## ABSTRACT

Carotid Intima-media thickness (CIMT) and plaque are well established markers of subclinical atherosclerosis and are widely used for identifying subclinical atherosclerotic disease. We performed association analyses using Metabochip array to identify genetic variants that influence variation in CIMT and plaque, measured using B-mode ultrasonography, in rheumatoid arthritis (RA) patients. Data on genetic associations of common variants associated with both CIMT and plaque in RA subjects involving Mexican Americans (MA) and European Americans (EA) populations are presented in this article. Strong associations were observed after adjusting for covariate effects including baseline clinical characteristics and statin use. Susceptibility loci and genes and/or nearest genes associated with CIMT in MAs and EAs with

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RA are presented. In addition, common susceptibility loci influencing CIMT and plaque in both MAs and EAs have been presented. Polygenic Risk Score (PRS) plots showing complementary evidence for the observed CIMT and plaque association signals are also shown in this article. For further interpretation and details, please see the research article titled “A Genetic Association Study of Carotid Intima-Media Thickness (CIMT) and Plaque in Mexican Americans and European Americans with Rheumatoid Arthritis” which is being published in *Atherosclerosis* (Arya et al., 2018) [1]. (Arya et al., in press) Thus, common variants in several genes exhibited significant associations with CIMT and plaque in both MAs and EAs as presented in this article. These findings may help understand the genetic architecture of subclinical atherosclerosis in RA populations.

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### Specifications table

Subject area	<i>Genetics, Genomics and Molecular Biology</i>
More specific subject area	<i>CIMT and plaque Genetic Association Data in RA subjects</i>
Type of data	<i>Tables, figures and text file</i>
How data was acquired	<i>ORALE study data were collected from 11 private and public rheumatology outpatient clinics in San Antonio, Texas.</i>
Data format	<i>Original and analyzed data set</i>
Experimental factors	<i>Recruited patients who met the 1987 criteria for RA</i>
Experimental features	<i>Metabochip array-based Association Analyses</i>
Data source location	<i>Division of Rheumatology and Clinical Immunology, Department of Medicine, UTHSCSA, San Antonio, TX USA.</i>
Data accessibility	<i>Data are available with this article and/or upon request</i>

### Value of the data.

- Genetic association data for CIMT and plaque in subjects with rheumatoid arthritis are valuable to understand the genetic architecture of the carotid atherosclerosis markers in RA patients.
- Observed association signals for carotid atherosclerosis markers in both Mexican American and European American cohorts would give more insight into population differences as well as trait-specific and common genetic determinants.
- This data could also be potentially used for replication of genetic association findings for atherosclerosis markers in other populations.

### 1. Data

A comprehensive association results of the transformed CIMT after adjustment for covariate effects in MAs and EAs are shown in [Tables 1](#) and [2](#), respectively. As presented in Arya et al. [1], a total of 24 SNPs from 11 chromosomes exhibited association with CIMT in MAs at  $p < 1 \times 10^{-4}$ , and the  $p$  values ranged from  $9.95 \times 10^{-5}$  to  $3.80 \times 10^{-6}$ , while 12 SNPs from 7 chromosomes were associated with

**Table 1**  
Susceptibility loci associated with carotid IMT in Mexican Americans.

Chr.	SNP	Position, bp <sup>a</sup>	Gene/nearest gene	Loc.	A1 <sup>b</sup>	MAF	BETA	SE	P <sup>c</sup>
6	rs17526722	26,026,834	SLC17A2	I	A	0.10	-0.8377	0.1797	3.80E-06
6	rs36014129	25,992,498	SLC17A3   SLC17A2	IG	A	0.10	-0.85	0.184	4.63E-06
6	rs13212534	26,090,989	TRIM38	I	A	0.10	-0.8607	0.1891	6.36E-06
6	rs13213957	25,894,205	SLC17A1	I	C	0.10	-0.7806	0.1762	1.10E-05
6	rs55912630	25,974,914	SLC17A3	I	C	0.10	-0.7806	0.1762	1.10E-05
6	rs13191296	25,792,585	SCGN	I	T	0.08	-0.7771	0.176	1.19E-05
2	rs4894108	180,112,792	ZNF385B	I	G	0.19	-0.2501	0.05751	1.59E-05
6	rs13211947	25,972,797	SLC17A3	I	T	0.10	-0.8179	0.189	1.75E-05
6	rs41266779	26,129,851	HIST1H3A   HIST1H4A	IG	T	0.10	-0.8178	0.1891	1.78E-05
17	rs2672901	76,411,261	KIAA1303	I	A	0.32	-0.1929	0.04489	2.00E-05
13	rs11619113	109,716,661	COL4A1	I	G	0.20	-0.3918	0.09165	2.21E-05
13	rs12873154	109,718,853	COL4A1	I	G	0.20	-0.3918	0.09165	2.21E-05
13	rs11619038	109,721,800	COL4A1	I	T	0.20	-0.3918	0.09165	2.21E-05
6	rs13202688	26,101,448	TRIM38   HIST1H1A	IG	G	0.10	-0.7505	0.1762	2.37E-05
10	rs61850526	63,190,012	C10orf107	I	T	0.02	-0.4624	0.1126	4.56E-05
6	rs34043431	25,983,063	SLC17A3   SLC17A2	IG	C	0.10	-0.7012	0.173	5.68E-05
11	rs76599700	61,401,283	FADS3	I	T	0.26	-0.7317	0.1814	6.17E-05
16	rs11860529	82,330,023	CDH13	I	T	0.08	0.4112	0.1028	7.06E-05
1	rs17436982	219,622,889	HLX   DUSP10	IG	T	0.22	0.2827	0.07173	8.98E-05
5	rs250216	50,317,115	PARP8   LOC642366	IG	C	0.07	0.2322	0.05899	9.20E-05
7	rs11761467	27,828,130	TAX1BP1	I	T	0.10	-0.3244	0.08256	9.44E-05
13	rs11620140	109,730,512	COL4A1	I	C	0.11	-0.3696	0.09417	9.62E-05
6	rs11966018	12,317,214	HIVEP1   EDN1	IG	C	0.03	0.691	0.1763	9.82E-05
15	rs7177074	97,357,475	LOC145814	I	A	0.05	-0.5138	0.1312	9.95E-05

Chr. = chromosome; SNP = single nucleotide polymorphism; Loc = Location; I = Intron; IG = Intergenic; MAF = minor allele frequency.

<sup>a</sup> Based on National Center for Biotechnology Information (NCBI) Build36 (hg18).

<sup>b</sup> A1 = minor allele.

<sup>c</sup> p values ranked from low to high.

**Table 2**  
Susceptibility loci associated with carotid IMT in European Americans.

Chr.	SNP	Position, bp <sup>a</sup>	Gene/nearest gene	Loc.	A1 <sup>b</sup>	MAF	BETA	SE	P <sup>c</sup>
15	rs1867148	73,127,038	PPCDC	I	C	0.44	-0.2804	0.06064	5.11E-06
15	rs7163636	73,133,813	PPCDC / C15orf39	IG	C	0.48	-0.2748	0.06053	7.51E-06
13	rs323453	104,232,379	LOC728183   DAOA	IG	G	0.34	-0.2724	0.06346	2.23E-05
15	rs3812945	73,076,775	SCAMPS	I	G	0.45	0.2626	0.06346	2.23E-05
1	rs4846566	217,797,888	LOC728510   ZC3H11B	IG	T	0.01	-1.186	0.2777	2.46E-05
2	rs12987042	38,518,674	ARLGP2 IRPLPO-like	IG	A	0.36	-0.2608	0.0619	3.13E-05
1	rs17006057	217,790,022	LOC728510/ZC3H11B	IG	A	0.01	-1.225	0.2915	3.29E-05
15	rs6495122	72,912,698	CPLX3   ULK3	IG	A	0.42	-0.2525	0.06149	4.90E-05
1	rs2645091	2,214,505	SKI	I	T	0.15	-0.3511	0.08741	7.07E-05
16	rs17821532	52,504,199	FTO	I	A	0.05	-0.7006	0.1751	7.58E-05
6	rs7742814	119,185,974	C6orf204 ASF1A	IG	G	0.36	0.2555	0.06462	9.09E-05
11	rs4387380	5,824,023	OR52E6 OR52E8	IG	C	0.03	-0.8226	0.2085	9.45E-05

Chr. = chromosome; SNP = single nucleotide polymorphism; Loc = Location; I = Intron; IG = Intergenic; MAF = minor allele frequency.

<sup>a</sup> Based on National Center for Biotechnology Information (NCBI) Build36 (hg18).

<sup>b</sup> A1 = minor allele.

<sup>c</sup> p values ranked from low to high.

**Table 3**

Common susceptibility loci associated with carotid IMT and plaque in Mexican Americans.

Chr.	SNP	Position, bp <sup>a</sup>	Gene/nearest gene	Loc.	A1 <sup>b</sup>	MAF	BETA	SE	P <sup>c</sup> CIMT	P <sup>c</sup> Plaque
6	rs17526722	26,026,834	SLC17A2	I	A	0.10	-0.8377	0.1797	3.80E-06	0.04112
2	rs4894108	180,112,792	ZNF385B	I	G	0.19	-0.2501	0.05751	1.59E-05	0.00823
17	rs2672901	76,411,261	KIAA1303	I	A	0.32	-0.1929	0.04489	2.00E-05	0.01672
13	rs12873154	109,718,853	COL4A1	I	G	0.20	-0.3918	0.09165	2.21E-05	0.04923
10	rs61850526	63,190,012	C10orf107	I	T	0.02	-0.4624	0.1126	4.56E-05	0.001367
11	rs76599700	61,401,283	FADS3	I	T	0.26	-0.7317	0.1814	6.17E-05	0.08027
16	rs11860529	82,330,023	CDH13	I	T	0.08	0.4112	0.1028	7.06E-05	0.02431
1	rs17436982	219,622,889	HLX   DUSP10	IG	T	0.22	0.2827	0.07173	8.98E-05	0.006092
5	rs250216	50,317,115	PARP8   LOC642366	IG	C	0.07	0.2322	0.05899	9.20E-05	0.0262
7	rs11761467	27,828,130	TAX1BP1	I	T	0.10	-0.3244	0.08256	9.44E-05	0.01331
6	rs11966018	12,317,214	HIVEP1   EDN1	IG	C	0.03	0.691	0.1763	9.82E-05	0.009176
15	rs7177074	97,357,475	LOC145814	I	A	0.05	-0.5138	0.1312	9.95E-05	0.001647

Chr. = chromosome; SNP = single nucleotide polymorphism; Loc = Location; I = Intron; IG = Intergenic; MAF = minor allele frequency.

<sup>a</sup> Based on National Center for Biotechnology Information (NCBI) Build36/130 (hg18).

<sup>b</sup> A1 = minor allele.

<sup>c</sup> *p* values ranked from low to high.

**Table 4**

Common susceptibility loci associated with plaque and CIMT in Mexican Americans.

Chr.	SNP	Position, bp <sup>a</sup>	Gene/nearest gene	Loc	A1 <sup>b</sup>	MAF	OR	SE	P <sup>c</sup> Plaque	P <sup>c</sup> CIMT
13	rs496916	109,649,015	COL4A1	I	C	0.41	0.514	0.1428	3.15E-06	0.03256
15	rs9806753	46,953,709	SHC4   EID1	IG	A	0.26	1.739	0.1229	6.71E-06	0.1274
6	rs9463110	12,890,588	PHACTR1	I	G	0.44	1.699	0.1222	1.45E-05	0.02229
22	rs2092179	38,364,539	CACNA11	I	C	0.25	0.5883	0.1289	3.85E-05	0.02903
9	rs7869506	98,127,033	SLC35D2	I	T	0.26	1.688	0.1296	5.41E-05	0.005388
1	rs6667860	36,730,800	CSF3RIGRIK3	IG	C	0.48	1.658	0.1298	9.76E-05	0.0002729

Chr. = chromosome; SNP = single nucleotide polymorphism; Loc = Location; I = Intron; IG = Intergenic; MAF = minor allele frequency.

<sup>a</sup> Based on National Center for Biotechnology Information (NCBI) Build36 (hg18); OR = Odds Ratio.

<sup>b</sup> A1 = minor allele.

<sup>c</sup> *p* values ranked from low to high.

CIMT in EAs and the *p* values ranged from  $9.45 \times 10^{-5}$  to  $5.11 \times 10^{-6}$ . The best associated SNPs are different in both populations.

Top SNPs that are associated with CIMT and plaque were different in significance levels but exhibited associations with either phenotype at nominal significance levels ( $p < 0.05$ ) as shown in Tables 3–6. It is well known that some variants exhibit unique associations with a given phenotype (CIMT or plaque) while other variants exhibit common associations with both phenotypes (CIMT and plaque). Despite the correlation between CIMT and plaque, different trait-specific genetic determinants can be expected due to their differing pathobiology and associated phenotypic severity of plaque as shown by our earlier study as well as other studies [2,3]. In addition, genetic differences are expected between the populations of European background and admixed populations such as the Mexican Americans that have both European and Native American ancestries. Furthermore, results from previous studies also support our findings [4,5].

As shown in Fig. 1, Quantile-Quantile (Q-Q) plots of the transformed CIMT and plaque in both MAS and EAs have been generated using EPACTS. Q-Q plots exhibited a roughly straight line through the origin with a unit slope indicating almost no inflation.

To further investigate the genetic architecture of CIMT, PRS analysis was conducted, with scores representing summations of CIMT-, and plaque-associated alleles from the MetaboChip array. A PRS is

**Table 5**

Common susceptibility loci associated with CIMT and plaque in European Americans.

Chr.	SNP	Position, bp <sup>a</sup>	Gene/Nearest Gene	Loc.	A1 <sup>b</sup>	MAF	BETA	SE	P <sup>c</sup> CIMT	P <sup>c</sup> plaque
15	rs1867148	73,127,038	PPCDC	I	C	0.44	-0.2804	0.06064	5.11E-06	0.01242
13	rs323453	104,232,379	LOC728183   DAOA	IG	G	0.34	-0.2724	0.06346	2.23E-05	0.2049
1	rs4846566	217,797,888	LOC728510   ZC3H11B	IG	T	0.01	-1.186	0.2777	2.46E-05	0.0676
2	rs12987042	38,518,674	ARL6IP2 RPLPO-like	IG	A	0.36	-0.2608	0.0619	3.13E-05	0.0005153
15	rs6495122	72,912,698	CPLX3   ULK3	IG	A	0.42	-0.2525	0.06149	4.90E-05	0.005717
1	rs2645091	2,214,505	SKI	I	T	0.15	-0.3511	0.08741	7.07E-05	0.03331
16	rs17821532	52,504,199	FTO	I	A	0.05	-0.7006	0.1751	7.58E-05	0.03347
6	rs7742814	119,185,974	C6orf204 ASF1A	IG	G	0.36	0.2555	0.06462	9.09E-05	0.3764
11	rs4387380	5,824,023	OR52E6 OR52E8	IG	C	0.03	-0.8226	0.2085	9.45E-05	0.06833

Chr. = chromosome; SNP = single nucleotide polymorphism; Loc = Location; I = Intron; IG = Intergenic; MAF = minor allele frequency.

<sup>a</sup> Based on National Center for Biotechnology Information (NCBI) Build36 (hg18).

<sup>b</sup> A1 = minor allele.

<sup>c</sup> p values ranked from low to high.

**Table 6**

Common susceptibility loci associated with carotid plaque and CIMT in European Americans.

Chr.	SNP	Position, bp <sup>a</sup>	Gene	Loc.	A1 <sup>b</sup>	MAF	OR	SE	P <sup>c</sup> Plaque	P <sup>c</sup> CIMT
12	rs515291	38,689,259	SLC2A13	I	G	0.32	0.4987	0.167	3.09E-05	0.0007138
11	rs10501399	68,979,039	MYEOVCCND1	IG	T	0.09	0.2938	0.2952	3.34E-05	0.001367
15	rs692390	39,274,282	EXDL1	I	A	0.19	0.4138	0.2157	4.30E-05	0.0001269
5	rs6887230	148,706,339	GRPEL2	I	G	0.30	2.02	0.1746	5.61E-05	0.009851
17	rs2070776	59,361,230	CD79B	C (ns)	T	0.33	0.5033	0.1722	6.68E-05	0.02914
16	rs3785233	7,607,511	A2BP1	I	C	0.16	0.419	0.2229	9.49E-05	0.0001409
6	rs10948573	50,800,310	TFAP2D	I	G	0.37	2.044	0.1834	9.68E-05	0.2071

Chr. = chromosome; SNP = single nucleotide polymorphism; Loc = Location; I = Intron; IG = Intergenic; MAF = minor allele frequency.

<sup>a</sup> Based on National Center for Biotechnology Information (NCBI) Build36 (hg18); OR = Odds Ratio.

<sup>b</sup> A1 = minor allele.

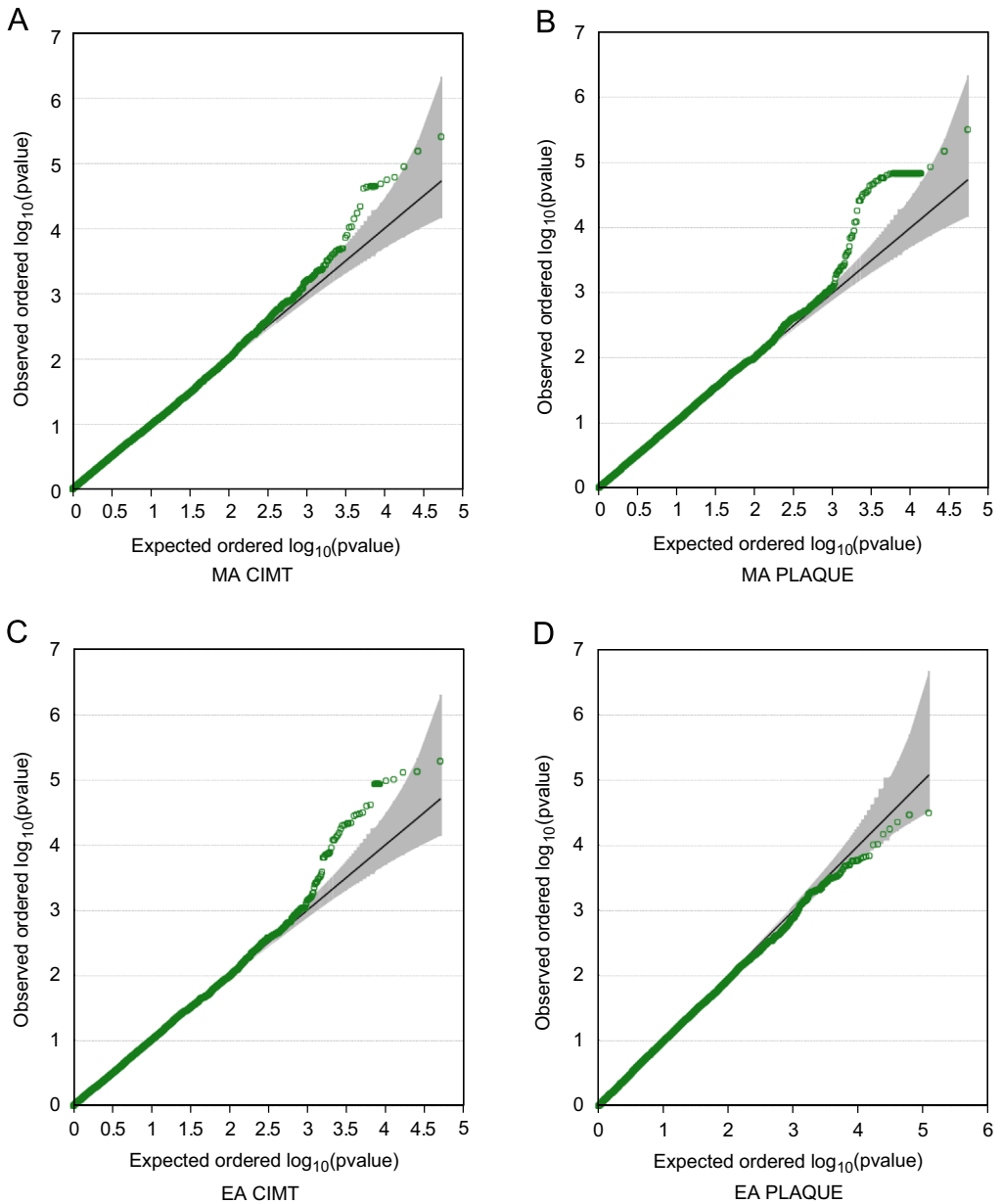
<sup>c</sup> p values ranked from low to high.

the sum of trait-associated alleles across many genetic loci, and was calculated in an independent population i.e., EA population, using the genome-wide association results from the MA population, a discovery population for a given trait (i.e., CIMT) to detect shared genetic etiology among traits, to establish the presence of a genetic signal in underpowered studies, and to infer the genetic architecture of a trait [6]. Scoring routines were determined from the association test results for the MA cohort, with risk alleles identified based on varying p-value thresholds (1,000 different p-value (Pts) thresholds, representing increments of  $p = 0.001$ ), each weighted by their estimated effect sizes on CIMT or plaque. Scores were then computed in the independent EA cohort, and evaluated as predictors of CIMT or plaque via regression models (covariates age, sex, PCs 1 and 2, RA duration, statin use, and htn) as shown in Fig. 2 (A and B). This work was performed using PRSice v.1.23, a polygenic risk score software.

## 2. Experimental design, subjects and methods

### 2.1. Subjects

We used existing samples/data from the ORALE (Outcome of Rheumatoid Arthritis Longitudinal Evaluation) study, involving 700 unrelated MAs and unrelated 415 EAs. From 1996 to 2009, we



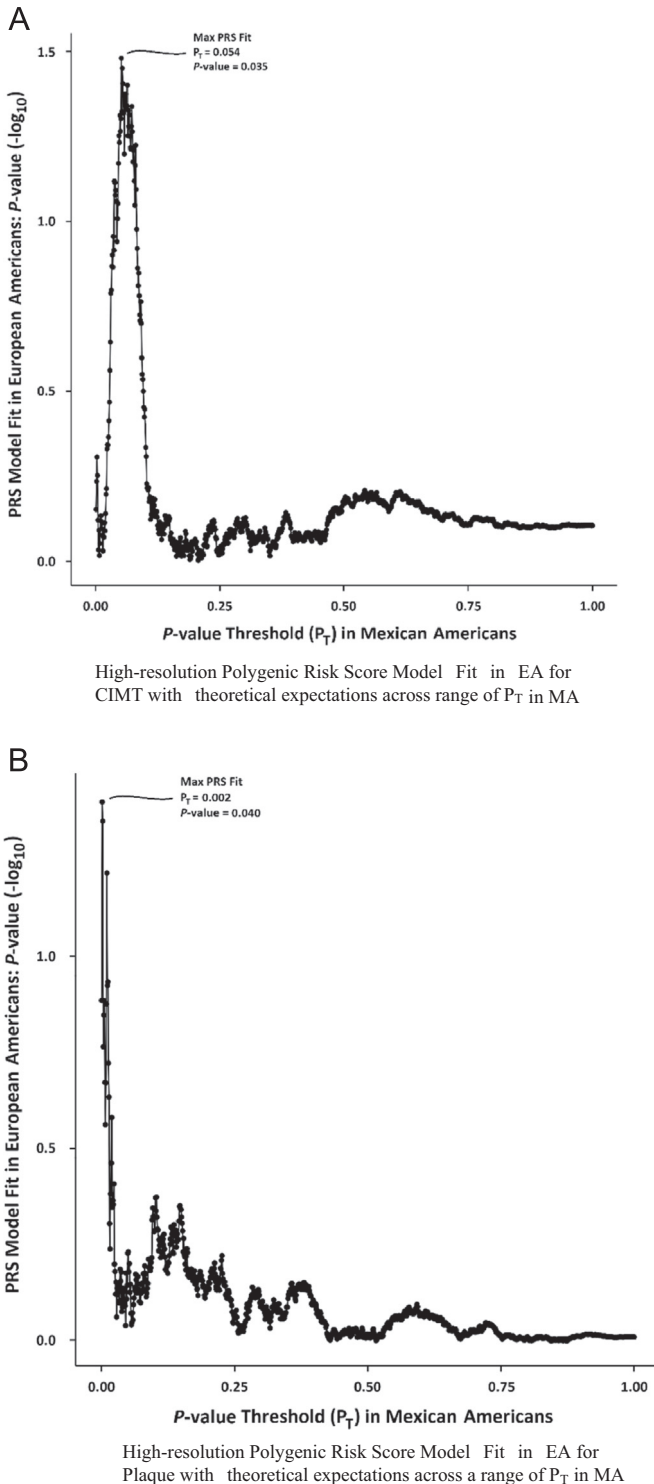
**Fig. 1.** Q-Q plots for CIMT and plaque association p-values in MAs and EAs.

recruited consecutive patients who met the 1987 criteria for RA [7] from 11 private and public rheumatology outpatient clinics in San Antonio, Texas.

## 2.2. Phenotyping

CIMT and plaque were measured using carotid ultrasound. A single technician performed a duplex scan of the carotid arteries in all patients, following a standardized vascular protocol as implemented





**Fig. 2.** High-resolution Polygenic Risk Score Model Fit in EA for CIMT (A) and Plaque (B) with theoretical expectations across a range of  $P_T$  in MA. A. High-resolution Polygenic Risk Score Model Fit in EA for CIMT with theoretical expectations across a range of  $P_T$  in MA. B. High-resolution Polygenic Risk Score Model Fit in EA for Plaque with theoretical expectations across a range of  $P_T$  in MA.

in an ATL HDI-3000 High Resolution Imaging machine with a L7-4 Transducer (Philips Medical Systems North America Company, Bothell, WA). For CIMT, we measured the end diastole at each of the near and far walls of the right and left common carotid arteries, and the anterior oblique, lateral and posterior oblique views of the internal carotid artery, for a total of 16 CIMT measurements per person. The maximal CIMT of the common and internal carotid arteries were obtained by averaging the maximal measurement from the near and far walls at each projection, from the right and left sides. Then the composite maximal CIMT was calculated by averaging the common and internal carotid maximal CIMT values. The result is one CIMT value per person, expressed in millimeters. Carotid plaque was identified as a discrete projection of 50% or more from the adjacent wall into the vessel lumen.

### 2.3. Genotyping

The Metabochip (Illumina) is a custom BeadChip targeting 196,725 genetic variants. Common and less common genetic variants were chosen from among the first iteration of the 1000 Genomes Project and represent index GWAS-identified variants regardless of disease or phenotype as of 2009 [8]. As previously described [8], it was primarily designed for fine mapping of metabolic and cardiovascular disease-related loci, and replication of susceptibility loci for specific GWAS-identified regions associated with cardio-metabolic disease and related phenotypes. Several studies have used this platform to successfully identify genetic risk factors influencing complex disease phenotypes [8,9].

Briefly, Metabochip was a custom Illumina iSelect genotyping array designed to test ~200,000 SNPs identified through genome-wide meta-analyses for metabolic and atherosclerotic/ cardiovascular diseases and traits in a cost-effective manner. It was designed by representatives of the following GWAS meta-analysis Consortia: CARDIoGRAM (coronary artery disease), DIAGRAM (type 2 diabetes), GIANT (height and weight), MAGIC (glycemic traits), Lipids (lipids), ICBP-GWAS (blood pressure), and QT-IGC (QT interval). It supports genotyping of SNPs selected according to five sets of criteria: (1) individual SNPs displaying evidence for association in GWA meta-analyses to diseases and traits relevant to metabolic and atherosclerotic-cardiovascular endpoints, (2) detailed fine mapping of loci validated at genome-wide significance from these meta-analyses, (3) all SNPs associated at genome-wide significance with any human trait, (4) "wildcards" selected by each meta-analysis Consortium for Consortium-specific purposes, and (5) other useful content, including SNPs that tag common CNPs, SNPs in the HLA region, SNPs marking the X and Y chromosomes and mtDNA, and for sample fingerprinting (common SNPs represented on major genome-wide array products from both Illumina and Affymetrix) [8,10]. After merging and pruning the lists (to remove redundant SNPs), a total of 217,697 SNPs representing 245,243 bead types was submitted to Illumina for manufacturing on August 19, 2009. The final chip consisted of genotypes of ~200,000 SNPs per sample. We performed the genotyping according to the Illumina protocol and initial data handling and analysis was performed using Genome Studio v1.7.4 (Illumina).

### 2.4. Sample and SNP quality control measures

Several quality control measures were applied to the genotypic data of each ethnic group, and only the autosomal SNPs that passed QC were considered for this study. Subjects with low call rates ( $< 0.95$ ) were removed ( $MA = 13$  and  $EA = 0$ ). To identify and exclude highly related individuals or duplicate samples, we performed the relationship inference analytical procedure as implemented in the computer program KING [Kinship-based Inference for Genome-wide association studies, [11] and identified related individuals. Subsequently, using the program PLINK [12] and the identity-by-descent (IBD) analysis, closely related individuals up to 3<sup>rd</sup> degree relatives ( $IBD > 0.185$ ) were removed ( $MA = 17$  and  $EA = 3$ ). To detect ethnic outliers, we used EIGENSTRAT c3.0 software package [13] to employ principal components analysis to a subset of autosomal SNPs in our data that were in low LD ( $r^2 < 0.2$ ) and the HapMap samples as reference for the ethnic groups. Plots were generated using the first two principal components (PCs) for visual inspection. Using our data by ethnic group, samples were identified as population outliers, defined by 4SD from the mean of each of the 2 PCs

that explained the majority of variation in the data, and were subsequently removed ( $MA = 2$  and  $EA = 0$ ). SNPs with a genotyping call rate less than 95% were removed using PLINK [12]. In addition, SNPs with Hardy-Weinberg Equilibrium (HWE) values of  $p < 10^{-4}$  [( $MA = 236$  (CIMT) and 120 (plaque); and  $EA = 114$  (CIMT), and 38 (plaque))] and with minor allele frequency (MAF)  $< 0.01$  ( $MA = 57, 323$  and  $EA = 60, 168$ ) were removed from the analysis. After filtering and genotyping pruning, 122,549 SNPs from 668 MAs and 120,827 from 415 EAs were remained in the association analyses.

### 2.5. Quantile-quantile (Q-Q) plots

Q-Q plots are probability plots, which are useful to compare two probability distributions, sample quantile distribution of the observed chi-squared values (y-axis) versus the quantile distribution of expected (normal or theoretical) chi-squared values (x-axis) graphically by plotting their quantiles against each other. Q-Q plots were done using Efficient and Parallelizable Association Container Toolbox software [EPACTS, <http://genome.sph.umich.edu/wiki/EPACTS>]. Association  $p$  values were adjusted for multiple testing using the conservative Bonferroni correction:  $4.08 \times 10^{-7}$  for MA and  $4.14 \times 10^{-7}$  for EA.

### 2.6. Statistical genetic analyses

We performed association analyses between the transformed CIMT (as a quantitative trait) and plaque (as a discrete trait) and SNP genotypes in both MA and EA samples after QCs, using PLINK software version 1.07 [12]. Principal Components (PCs) were derived using EIGENSTRAT principal component analysis [13] to adjust for potential population stratification influences. A linear regression additive genetic model (SNPs coded as 0,1, or 2 based on the minor allele dosage) adjusted for the effects of covariates age, sex, RA duration, medication status (statin use, and hypertension, [htn, medication]), and the first two PC1 and PC2, was used for association testing of CIMT, a quantitative trait. Association statistics for plaque, a discrete trait, were calculated using logistic regression assuming an additive model.

### 2.7. Polygenic risk score (PRS)

PRS for an individual is a summation of their genotypes at variants genome-wide, weighted by effect sizes on a trait of interest i.e. CIMT. Although effect sizes are usually estimated from a GWAS study, we used our Mexican American cohort association results for weighting. Thus, a sum of trait-associated alleles across many genetic loci, has been calculated in an independent population i.e., European American population, using the genome-wide association results from the Mexican-American population, a discovery population for a given trait (i.e., CIMT) to detect shared genetic etiology among traits, to establish the presence of a genetic signal in underpowered studies, and to infer the genetic architecture of a trait [6].

Using the array-wide association results for the MA samples for CIMT ( $n = 122,549$  SNPs), PRS routines were designed and computed in the EA cohort, revealing significant associations with the target phenotype. As shown in Figs. 1 and 2, the PRS model for  $P_t < 0.054$  yielded the best fit for CIMT in EAs, but 1000 different  $P_t$  thresholds were examined, representing increments of  $P = 0.001$ . The computed PRSs are then used as predictors of a targeted phenotype in an independent European data set using regression models (i.e., linear or logistic based on the target phenotype). Furthermore, SNPs in the association results for the Mexican-American samples were pruned using PLINK's clumping methodology based on linkage disequilibrium (LD), distance, and association P-values (see <http://pngu.mgh.harvard.edu/~purcell/plink/clump.shtml>). We used the standard settings ( $r^2 = 0.1$  and 250 Kb), which reduced the number of SNPs actually used in the scoring routines from 122,549 to 36,630. This work was performed using PRSice v.1.23, a polygenic risk score software [6].

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## Transparency document. Supporting information

Supplementary data associated with this article can be found in the online version at [doi:10.1016/j.dib.2018.02.006](https://doi.org/10.1016/j.dib.2018.02.006).

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