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Gene-based evaluation of low frequency variation and genetically predicted gene expression impacting risk of keloid formation

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SUMMARY

Keloids are benign dermal tumors occurring ~20 times more often in African- as compared to European-descent individuals. While most keloids occur sporadically, a genetic predisposition is supported by both familial aggregation of some keloids and large differences in risk among populations. Despite Africans and African Americans being at increased risk over lighter-skinned individuals, little genetic research exists into this phenotype. We reported, using a combination of admixture mapping and exome analysis, multiple common variants within chr15q21.2-22.3 associated with risk of keloid formation in African Americans. Here we describe a gene-based association analysis using 478 African American samples with exome genotyping data to identify genes containing low-frequency variants associated with keloids, with evaluation of genetically predicted gene expression in skin tissues using association summary statistics. The strongest signal from gene-based association was located in C15orf63 (p-value = 6.6×10^{-6}) located at 15q15.3. The top result from gene expression was increased predicted DCAF4 expression (p-value = 5.5×10^{-4}) in non-sun-exposed skin, followed by increased predicted *OR10A3* expression in sunexposed skin (p-value = 6.9×10^{-4}). Our findings identify variation with putative role in keloid formation, enhanced by the use of predicted gene expression to support the biological roles of variation identified only though genetic association studies.

Conflict of Interest Statement: The authors declare no conflicts of interest

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Keywords

African Americans; gene expression; genetic predisposition; keloid; skin

INTRODUCTION

Keloids are benign dermal tumors that form during prolonged wound healing as the result of a fibroproliferative process(Marneros & Krieg, 2004, Niessen *et al.*, 1999). Although some cases of keloid formation may be due to somatic mutation(Saed *et al.*, 1998, Ladin *et al.*, 1998), multiple keloids in the same individual and evidence for a multicellular origin of keloids(Chevray & Manson, 2004, Moulton-Levy *et al.*, 1984) argue against somatic mutation as the cause in most cases. Several lines of evidence support a genetic basis for keloids, including the occurrence of familial forms and racial and ethnic differences in prevalence(He *et al.*, 2017). Keloid scarring is more prevalent in darker-skinned individuals(Ud-Din & Bayat, 2013) with approximately 20-fold higher prevalence in African Americans (AAs) compared to European Americans (EAs) in the US(Barrett, 1973).

Several studies have attempted to identify the genetic bases of keloids, but few associating genes have been identified (Brown *et al.*, 2008, Halim *et al.*, 2012, Shih & Bayat, 2010, Nakashima *et al.*, 2010, Emami *et al.*, 2012, Yan *et al.*, 2007, Chen *et al.*, 2007, Marneros *et al.*, 2004). However, in the only published genome-wide association study (GWAS) of keloids, in a Japanese population(Nakashima *et al.*, 2010), four single nucleotide polymorphisms (SNPs) at three loci (1q41, 3q22.3-23 and 15q21.3) showed significant association with keloid formation; two of these three (1q41 and 15q21.3) replicated in a Chinese population(Zhu *et al.*, 2013). Using admixture mapping our group identified a genomic region on chr15q21.2-22.3 with increased local African ancestry that associated with keloid risk and that included multiple common genetic variants impacting keloid formation in AAs(Velez Edwards *et al.*, 2014). The region identified through admixture mapping included *NEDD4*, a gene implicated in the prior Japanese keloid GWAS. However, the strongest associations were in a nearby gene, *MYO1E*.

Given the success of gene-based analyses for identifying genes harboring multiple rare variants associating with disease(Wessel *et al.*, 2015, Huyghe *et al.*, 2013, Do *et al.*, 2015, Cirulli, 2016), we sought to extend our previous association analysis into less common genetic variants through the use of gene-based regression and genetically predicted gene expression (GPGE) in (non-keloid) skin tissues from GTEx(2015). This latter analysis can provide *in silico* supporting evidence by which genetic variation may impact the phenotype in a tissue-specific manner, particularly when functional studies in the samples of interest are difficult or impossible to perform.

METHODS

Sample Collection and Phenotyping

Samples used in this study have been previously described(Velez Edwards *et al.*, 2014). These studies were approved by the Vanderbilt University Institutional Review Board.

Briefly, the study included DNA from 71 AA keloid cases and 399 controls from the BioVU DNA Repository and 36 cases from the keloid fibroblast repository described in detail below. BioVU keloid cases were defined as AAs 18 years or older who were diagnosed with keloids in the electronic health record (EHR) of patients at Vanderbilt University Medical Center (VUMC) and who have at least two mentions of a keloid diagnosis in their record (either two diagnostic codes [International Classification of Diseases (ICD) version 9 = 701.4] or a code and mention of a keloid within their record).

Keloid cases (n=36) were also obtained from a repository (15-53 year olds) that includes cultured fibroblasts from normal and keloid scar tissue(Russell *et al.*, 2010, Smith *et al.*, 2008). The diagnosis was made both by the surgeon or dermatologist removing the tissue and by the pathologist who examined the tissue. The principal criterion used to differentiate keloid from other hypertrophic scars was the extent to which the scar exceeded the boundary of the initiatial wound. DNA was also obtained from another repository of blood samples of unrelated individuals (n = 21) who were part of multiplex families.

Controls were AA subjects 18 years and older who have had surgical procedures performed at VUMC that involved an open wound, such as breast surgery, cesarean section and open heart surgery, and have two years of follow-up in the EHR with no evidence of keloid formation. Controls were excluded if they had ICD-9 codes for other fibroproliferative diseases (asthma, nephrosclerosis, or fibroids) in their EHR, or if the words "excessive scarring" were present.

Genotyping

Genotyping and quality control procedures have also been described elsewhere(Velez Edwards *et al.*, 2014). Briefly, DNA samples were isolated from whole blood using the Autopure LS system (QIAGEN Inc., Valencia, CA). We genotyped DNA from the 492 participants using the custom Affymetrix Axiom Exome Genotyping Array (Affymetrix Inc., Santa Clara, CA). The genomic DNA samples were processed according to standard Affymetrix procedures for processing of the assay and genotype calling was performed using the Affymetrix Power Tools software (APT, Affymetrix Inc., Santa Clara, CA).

Genotyping Quality Control

Quality control procedures included evaluation of all SNPs for deviation from Hardy– Weinberg equilibrium (HWE) using PLINK software (Purcell et al. 2007). SNPs with HWE $p = 1.0 \times 10^{-6}$, low genotyping efficiency (<95 %), duplicates, non-autosomal locations, and those which were monomorphic were removed. After removal of subjects and SNPs for quality control, 478 subjects (122 cases and 356 controls) and 163,613 SNPs remained for analyses.

Statistical Analysis

The program EPACTS (Efficient and Parallelizable Association Container Toolbox; http:// csg.sph.umich.edu/kang/epacts/index.html) was used for variant annotation (ANNOVAR). This annotation grouped the variants by gene for use in gene-based burden tests. The Optimal Sequence Kernel Association Test (SKAT-O) was also implemented in EPACTS for

gene-based association among nonsynonymous (missense, nonsense and splice-site changes) variants. All nonsynonymous variants were used, with no allele frequency threshold for inclusion, as SKAT-O weights those variants with a minor allele frequency of less than 0.05 more heavily than common variation. Sex and 10 principal components to account for population stratification among the samples were included as covariates. The Bonferroni significance threshold for this analysis (p-value $< 3.9 \times 10^{-6}$) accounted for the number of tests, represented by the number of genes with more than one variant (N = 12,714)

In order to evaluate the genetic association results in the context of gene expression relevant to keloid scars, we employed the method S-PrediXcan(Barbeira et al., 2016), an extension of the PrediXcan method(Gamazon et al., 2015). This approach utilizes genetic association results in conjunction with GPGE levels to infer gene expression association with keloids. Briefly, all genetic variants within 1 MB of each gene are assessed to identify SNPs with an impact on that gene's expression (cis-eQTL SNPs). Accounting for linkage disequilibrium, SNPs are weighted for their relative effect on gene expression and collapsed using LASSO into a single predictive model for each gene's GPGE in each tissue. S-PrediXcan builds on this approach through the use of summary statistics, rather than individual level data. Given that each SNP also has statistics for their association with the phenotype of interest, the association between the GPGE levels and the phenotype (keloids) can be inferred. S-PrediXcan identifies regulatory mechanisms through which genetic variants affect phenotype (and the direction in which they do so) on the gene-level rather than the SNP, while largely avoiding reverse causality since predicted expression levels are based on germline variation, which are not affected by onset of disease. For the purposes of this study, we utilized predicted expression weights built for two tissues from GTEx: sun exposed skin from the lower leg and skin which was not sun-exposed (from the suprapubic region), as well as transformed fibroblast cells. The prediction models and covariance matrices used are available on PredictDB (http://predictdb.hakyimlab.org/). Gene prediction models were filtered to exclude those not passing a false discovery rate threshold of 5%. P-values < 0.05from more than one analysis were considered to be suggestive.

RESULTS

The majority of subjects were female (70% cases and 62% controls). The mean age of study participants was 43 ± 17 for cases and 52 ± 18 controls.

Gene-based association analyses

We identified 12,714 genes in the exome chip data with at least two variants meeting the annotation criteria. The SKAT-O association results are summarized in Figure 1. The top result, though not reaching conservative statistical significance for the number of tests, was the open reading frame *C15orf63* (p-value = 6.61×10^{-6}), which contained two low-frequency nonsynonymous variants, one of which was seen only once (singleton). A total of 12 variants reached suggestive levels of significance (p-value < 1×10^{-4}), these are presented in Table 2.

Our previous admixture analysis and single-variant associations revealed a region of interest on chromosome 15, using common variants from the same exome chip. This extension of

that work also had a suggestive hit on chromosome 15. However, when examining these SKAT-O results (which emphasize the uncommon variation) in the context of the admixture mapping, the top signals on the chromosome are proximal to the admixture mapping peak (Figure 2; Table 3). Further, the genes previously implicated (*MYO1E* in our previous work, and *NEDD1* in a GWAS in Japanese subjects) are not associated in this gene-based analysis (p-value = 0.46 and 0.48, respectively), suggesting that the previously observed association at these genes are probably not due to known rare variants in those genes, but may be due to previously uncharacterized novel variants. Other genes reaching nominal significance in the admixture analyses from chromosome 15 are presented in Table 3. Nominally significant p-values for genes in this admixture mapping region ranged from <0.01 to 0.04.

Evaluation of gene expression using GTEx

In order to evaluate the genetic association results in the context of gene expression relevant to keloid scars, we applied S-PrediXcan method to the single variant summary statistics. S-PrediXcan can identify regulatory mechanisms through which genetic variants effect the phenotype (and the direction in which they do so) on the gene-level rather than the SNP, while largely avoiding reverse causality since predicted expression levels are based on germline variation, which are not affected by onset of disease. GPGE in two skin tissues from GTEx was evaluated using the summary statistics from the single variant association analysis of keloids (Figure 3). The most significant result was with increased predicted *DCAF4* expression in non-sun exposed skin tissue (from the suprapubic region; Table 3; p-value = 5.5×10^{-4}). This gene is located on chromosome 14. The maximal GPGE result in sun-exposed skin (from the lower leg) was with increased *OR10A3* on chromosome 11 (Table 3; p-value = 6.86×10^{-4}).

Overall, 23 genes had nominally significant (p<0.05) GPGE in both sun exposed and nonsun exposed skin, 10 from non-sun exposed tissue and 13 from sun exposed. There was no overlap of genes from the two tissues. Three of the 23 genes were also nominally associated through SKAT-O analysis of low frequency nonsynonymous variants (Table 4): *SLPI* and *ZNF337* on chromosome 20, and *ARFIP1* on chromosome 4. Most striking of these is *SLPI* (sun exposed p-value = 0.003, non-sun exposed p-value = 0.0392, SKAT-O p-value = 0.004), located on chromosome 20. Interestingly, increased predicted expression was associated with keloid risk in sun exposed skin, while decreased predicted expression associated with risk in non-sun exposed skin.

A small number of the nominally significant GPGE results in either of the two skin tissues were located on chromosome 15, where the top SKAT-O signal and previously identified admixture mapping signal are located. *FAM154B* and *RPAP1* were nominally (p-value < 0.05) associated in both sun-exposed skin and fibroblasts (data not shown), but not in nonsun exposed skin. Additional evaluation of the nominally significant results from SKAT-O the gene-based association testing revealed additional nominally significant GPGE results (Table 4).

DISCUSSION

This study sought to evaluate gene-based associations with keloids in AA through two mechanisms: the low-frequency coding variations assessed collectively for association with keloid risk, and prediction of keloid-associated genetic regulation of gene expression levels. The top result from the gene-based association tests was located in an open reading frame on chromosome 15. *C15orf63*, also known as *HYPK* (huntingtin interacting protein K), is a chaperone protein which interacts with huntingtin, the protein causing Huntington's disease(Choudhury & Bhattacharyya, 2015, Raychaudhuri *et al.*, 2014, Sakurai *et al.*, 2014).

We also employed S-PrediXcan to evaluate the association of the genetic variants implicated in keloid risk with expression in skin tissues from GTEx (Figure 3). This technique summarizes all eQTL variants (within one megabase [MB] of the gene) impacting a gene's expression into a single unit to infer association between GPGE in a given tissue and the outcome of interest. The most strongly associated GPGE with keloids was with increased *DCAF4* in non-sun exposed skin tissue (Table 3; p-value = 5.5×10^{-4}). This gene, DDB1 and CUL4 associated factor 4, is located on chromosome 14 and has been previously associated with leukocyte telomere length(Mangino *et al.*, 2015) and lung cancer risk(Liu *et al.*, 2017, Yan *et al.*, 2017). Notably, the variants associated with telomere length were also eQTLs in sun exposed skin, while there was no association between keloid risk and predicted expression of *DCAF4* in sun exposed skin in this study (p-value = 0.57).

Among the genes identified from gene-based association analyses, three had evidence from both skin tissues' gene expression as well as gene-based burden tests (Table 4). Most notable of these is SLPI (secretory leukocyte peptidase inhibitor), for which evidence supports a role in several skin disorders(Schafer et al., 2014, Lancto et al., 2013, Ashcroft et al., 2012, Skrzeczynska-Moncznik et al., 2012, Meyer-Hoffert, 2009, Bando et al., 2007), as well as a wide variety of other fibrotic diseases (Habgood et al., 2016, Hentschel et al., 2015, Nair et al., 2013, Aozasa et al., 2012, Thijs et al., 2015) and cancer(Zheng et al., 2016, Noorlag et al., 2015, Zuo et al., 2015, Rosso et al., 2014, Timms et al., 2014). SLPI has also been found to reduce contractility of fibroblast-mediated collagen gel models of scarring, suggesting that it may have uses in promotion of scarless wound healing(Sumi et al., 2000). The discordant directions of effect for SLPI predicted expression is interesting (Table 4), however this may reflect differences in skin where keloids are more likely to occur. However, as keloids are more likely to be found on the head/neck and upper extremities, neither the sun-exposed skin from the lower leg nor non-sun exposed skin from the suprapubic region can provide a clear picture of the most appropriate direction for keloid risk. ZNF337 (zinc finger protein 337) and ARFIP1 (ADP ribosylation factor interacting protein 1) were also nominally associated in sun-exposed and non-sun exposed skin and SKAT-O. Both of these genes are expressed in a wide variety of tissues from GTEx, but neither has yet been implicated in disease pathophysiology.

Although these gene-based analyses did not support the previous findings with *NEDD4* and *MYO1E* (Supplemental Table 1), it is important to note that those genes were implicated through mechanisms that emphasized the role of common variation (MAF>0.05), while the current study weighted only those uncommon variants that also were predicted to be

functionally important. Furthermore, the GPGE analysis was unable to construct models for either *NEDD4* or *MYO1E* GPGE. Therefore, we suggest that although common variation in *NEDD4* and *MYO1E* may be important for the development of keloid scars, rare variation in these genes does not appear to have an impact, and the relationship between the genetic association with keloids and gene expression levels of these genes in lower leg or suprapubic skin or fibroblast tissues remains undetermined.

Neither the gene-based association nor GPGE analyses reached multiple testing-corrected significance; however, considering convergent results which were suggestive in both analyses revealed genes with potential biological relevance to keloid risk. The lack of striking evidence for association in any one analysis may be due to small sample size and lack of comprehensive genomic evaluation. The association analyses presented here were limited to those known coding variants contained on the exome chip, which is known to be sub-optimal for non-EA populations(Nievergelt et al., 2014). The relatively small number of available SNPs also impacts the S-PrediXcan analysis, with only small proportions of the SNPs used in construction of the models available. Despite these challenges, studies of this type add new evidence to the body of literature evaluating the genetic component involved in the formation of keloid scarring in African American populations. Studies of genome-wide association in larger numbers of samples are necessary to refine the impact of genetic variation underlying keloid development across many population groups. This may be enabled by the growth of large biorepositories with genome-wide genotyping resources and linked to EHRs for ease of identification of cases and controls. Further knowledge may also be generated by obtaining eQTL databases for skin tissues more directly relevant to keloid formation, i.e. those located in chest/abdominal areas or face(Yedomon et al., 2012).

In summary, we evaluated whole-exome genotyping data for evidence of gene-based association with risk of keloids and observed modest evidence of association in several genes, based on both genetic and expression data. The finding that predicted expression of *SLPI* in skin tissue was associated with keloids is the first evidence of this type and is supported by the association with low frequency coding variants in that gene, as well as the known biology of this gene in skin disorders and potential role in scar formation. The results of this study indicate that the association of previously implicated genes *NEDD4* and *MYO1E* is not due to underlying associations with low-frequency or rare coding variants, at least in this sample of AAs, though further evaluation of these and other genes in diverse populations will continue to inform. Finally, the identification of multiple genes associated with previously known pathways suggests a plausible role for genetic variants impacting keloid risk in AAs.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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Figure 1. Summary of SKAT-O association results for keloids using exome chip variants Red line represents Bonferroni significance threshold for multiple testing correction.

Chr15 SKAT / Admixture Mapping



NEDD4 MYO1E

Figure 2. Focused summary of SKAT-O gene-based association results on chromosome 15 overlaid on admixture mapping results

SKAT-O results for formerly detected genes of interest *MYO1E* and *NEDD4* are indicated by green and red circles, respectively. Suggestive associations in this region flank the admixture mapping signal rather than occurring within the implicated region.

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Figure 3. S-PrediXcan results in three relevant tissues opposed from exome chip association results

Upper panel displays genetically predicted gene expression results for sun-exposed and nonsun exposed skin, and transformed fibroblasts, while the bottom panel shows the genetic association results for keloid risk.

Table 1

Summary of demographic characteristics and ancestry estimates

	Keloid Cases (N = 122)	Controls (N = 356)	p-value
Age ^a , years (Mean(SD))	43 (17)	52 (18)	>0.0001
Sex (% Female)	70	62	0.13
European Ancestry (%)	19	21	0.011

 a Age was only available for samples obtained from BioVU

Table 2

SKAT Results from exome-wide analyses.

Chr.	Position (hg19)	Band	Gene	Fraction Rare †	N SNPs	N Singletons	p-value
15	44092839-44093927	15q15.3	C15orf63	0.08	2	1	$6.41{\times}10^{-6}$
19	11350326-11350935	19p13.2	C19orf80	0.06	3	0	2.17×10^{-5}
2	85570442-85578987	2p11.2	RETSAT	0.07	8	2	2.22×10 ⁻⁵
1	54417796-54433454	1p32.3	LRRC42	0.02	2	0	3.41×10^{-5}
2	232457693-232458648	2q37.1	C2orf57	0.13	10	4	4.92×10^{-5}
2	23977086-24090746	2p24.1-23.3	ATAD2B	0.08	8	1	5.57×10^{-5}
10	90582744-90591710	10q23.31	ANKRD22	0.006	2	1	$6.94{\times}10^{-5}$
1	246727700–246729395	1q44	TFB2M	0.008	3	2	7.62×10 ⁻⁵
13	20567666–20637020	13q12.11	ZMAMZ	0.07	4	2	8.01×10^{-5}
10	106074366-106075500	10q25.1	ITPRIP	0.03	L	4	8.26×10 ⁻⁵
6	85863069–86153103	9q21.32	FRMD3	0.03	5	0	8.31×10^{-5}
7	102937991-102952123	7q22.1	PMPCB	0.07	9	2	9.79×10 ⁻⁵

 $\dot{\tau}^{}_{\rm Fraction}$ of Samples w/a Rare Variant

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SKAT N ^{‡‡}	10	8		3	1	2		1		2	3	11	2	9	2	1	4	3	10			4	
SKAT p-value	0.242	0.150	NA	0.424	0.341	0.769	NA	0.253	NA	0.528	0.103	0.279	1	0.052	0.135	0.004	0.733	1	0.336	NA	NA	0.603	
Model N $^{\dagger \dot{\tau}}$	33	45	43	25	61	44	36	37	18	61	29	41	62	5	11	47	82	44	14	43	31	105	
N Variants¶	2	1	1	1	2	1	2	1	1	1	1	2	2	1	1	1	1	3	1	1	1	3	
Q	0.002	0.018	0.042	1.73×10^{-4}	$6.00{\times}10^{-13}$	0.032	0.034	1.00×10^{-4}	0.002	$6.89{ imes}10^{-7}$	5.31×10^{-5}	$2.01{ imes}10^{-8}$	0.001	7.32×10 ⁻⁴	0.008	$3.78{\times}10^{-16}$	1.82×10^{-5}	3.32×10^{-11}	0.011	0.014	$4.68{ imes}10^{-9}$	9.65×10^{-6}	-
$\mathbf{R}^{2\sharp}$	0.045	0.012	0.008	0.043	0.249	0.020	0.009	0.047	0.051	0.079	0.086	0.101	0.052	0.034	0.035	0.204	0.096	0.140	0.016	0.028	0.121	0.102	
p-value	$5.52{\times}10^{-4}$	6.86×10^{-4}	0.001	0.002	0.002	0.002	0.002	0.002	0.002	0.003	0.002	0.003	0.002	0.003	0.003	0.003	0.003	0.003	0.004	0.003	0.004	0.004	
z-score †	3.45	3.40	3.22	-3.18	3.09	-3.08	-3.08	3.06	-3.03	-3.03	3.03	-3.01	2.99	2.98	2.97	2.96	2.95	2.94	-2.91	-2.91	-2.89	-2.86	
Skin Tissue	Non sun-exposed	Sun exposed	Sun exposed	Sun exposed	Non sun-exposed	Non sun-exposed	Sun exposed	Sun exposed	Non sun-exposed	Sun exposed	Non sun-exposed	Sun exposed	Non sun-exposed	Sun exposed	Non sun-exposed	Sun exposed	Non sun-exposed	Sun exposed	Sun exposed		Sun exposed	Non sun-exposed	
Band	14q24.2	11p15.4	7q31.31	10q24.33	19q13.42	1p13.3	14q11.2	14q24.1	2q33.1	1q24.3	7q22.1	17q25.1	20p11.21	19q13.33	1q22	20q13.12	19q13.41	19q13.43	17p13.1	17p13.1	19p13.12	3p25.3	
Position	73404680–73425478	7960139-7961010	117864712-117882784	105642542-105651971	56153418-56154836	110231746-110232965	21567096-21571883	67940444-67940444	198364721-198368187	172410967-172411374	100084515-100091399	74273284-74300497	25596850-25597061	50364549-50370425	156011427-156018299	43881680-43881682	52537202-52538294	57027720-57037309	8131548-8151342	8294023-8301144	14625576-14640134	10916706-10976854	
Chr	14	11	7	10	19	1	14	14	2	1	7	17	20	19	1	20	19	19	17	17	19	3	
Gene	DCAF4	OR10A3	ANKRD7	OBFCI	ZNF580	GSTMI	TMEM253	TMEM229B	HSPEI	PIGC	NYAPI	QRICH2	NANP	PNKP	UBQLN4	SLPI	ZNF432	ZNF471	CTCI	RNF222	DNAJBI	SLC6A11	*

Z-score for association of predicted gene expression with increasing risk of keloids. Direction of effect of the expression is indicated by sign of the z-score-

 $\sharp_{\rm Performance}$ prediction ${\rm R}^2$

 \S^{f} Prediction performance q-value.

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 π Mumber of SNPs included in the prediction model for that gene available in the summary statistics.

 $\dot{\tau}^{\star}_{N}$ Number of SNPs used to construct the prediction model for the gene in the tissue of interest using the GTEx data.

 $\frac{1}{2}$ Number of SNPs within the gene included in the SKAT analysis. NA = gene was not available in this analysis

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Table 4

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	5	3			Sun-ex	posed Skin		Non-Sun-	exposed Skin	SKAI	
Gene	Chr	Position	Band	z-score	p-value	N Variants/Model N †	z-score	p-value	N Variants/Model N †	p-value	Š.
IdTS	20	43881680-43881682	20q13.12	2.96	0.003	1/47	-2.06	0.039	3/77	0.004	1
ZNF337	20	25655765-25666737	20p11.1	-2.65	0.008	1/82	-2.65	0.008	1/28	0.032	4
ARFIPI	4	153791955–153802203	4q31.3	-2.21	0.027	1/15	-2.21	0.027	1/14	0.027	1
TMEM8B	6	35829222-35854844	9p13.3	-2.53	0.011	1/15	NA	NA	ΥN	0.004	5
TOR3A	1	179051112-179065129	1q25.2	-2.25	0.025	1/20	NA	NA	ΥN	0.043	2
MLXIPL	7	73010191–73020440	7q11.23	-2.62	0.00	3/16	-2.76	0.006	5/48	0.679	6
NAAA	4	76836110-76862005	4q21.1	2.73	0.006	4/46	2.63	0.00	3/31	0.263	8
ALDH16A1	19	49956613-49973578	19q13.33	-2.70	0.007	3/21	-2.30	0.022	2/58	0.423	14
CDRT4	17	15341161–15341332	17p12	2.56	0.010	2/23	2.67	0.008	1/31	0.234	13
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Boldface represents p-values < 0.05.

 $\dot{\tau}$ Number of SNPs included in the prediction model for that gene available in the summary statistics/Number of SNPs used to construct the prediction model for the gene in the tissue of interest using the GTEx data.

fNumber of SNPs with frequency <0.05 in the gene of interest and included in the analysis. NA = gene was not available in this analysis