

1 **A GWAS of ACE Inhibitor-Induced Angioedema in a South African Population.**

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29 **Abstract**

30 **Background:** Angiotensin-converting enzyme inhibitor-induced angioedema (AE-ACEI) is a
31 life-threatening adverse event and, globally, the commonest cause of emergency
32 presentations with angioedema. Several large genome-wide association studies (GWAS)
33 have found genomic associations with AE-ACEI. However, despite African Americans having
34 a 5-fold increased risk of AE-ACEI, there are no published GWAS from Africa. The aim of this
35 study was to conduct a case-control GWAS of AE-ACEI in a South African population and
36 perform a meta-analysis with an African American and European American population.

37 **Methods:** The GWAS included 202 South African adults with a history of AE-ACEI and 513
38 controls without angioedema following angiotensin-converting enzyme inhibitor (ACEI)
39 treatment for at least 2 years. A meta-analysis was conducted with GWAS summary
40 statistics from an African American and European American cohort (from
41 Vanderbilt/Marshfield with 174 cases and 489 controls).

42 **Results:** No SNPs attained genome-wide significance. However, 26 SNPs in the post-
43 imputation standard GWAS of the South African cohort and 37 SNPs in the meta-analysis
44 were associated to AE-ACEI with suggestive threshold ($p\text{-value} < 5.0 \times 10^{-06}$). Some of these
45 SNPs were found to be located close to the genes *PRKCQ* and *RIMS1*, previously linked with

46 drug-induced angioedema, and also close to the *CSMD1* gene linked to ACEI cough,
47 providing replication at the gene level, but with novel lead SNPs.

48 **Conclusions:** Our results highlight the importance of African populations to detect novel
49 variants in replication studies. Further increased sampling across the continent and matched
50 functional work are needed to confirm the importance of genetic variation in understanding
51 the biology of AE-ACEI.

52 **Keywords:** Angioedema, Genome-wide association studies, Angiotensin-converting Enzyme
53 Inhibitor.

54

55 **Non-standard Abbreviations and Acronyms:**

56 AE-ACEI: Angiotensin-converting Enzyme Inhibitor-Induced Angioedema

57 ACE: Angiotensin converting enzyme

58 ACEI: Angiotensin converting enzyme inhibitors

59 ARBs: Angiotensin receptor blockers

60 AGR: African Genome Resource

61 AWI-Gen: Africa Wits INDEPTH Partnership for Genomic Research

62 CREOLE: Comparison of three combination therapies in lowering blood pressure in Black
63 Africans

64 GCTA: Genome-wide complex trait analysis

65 GWAS: Genome-wide association studies

66 H3Africa: Human heredity and health in Africa

67 LMICs: Low-middle income countries

68 MAGMA: Multimarker analysis of genomic annotation

69 MLMA: Mixed linear model's association

70 QQ: Quantile-quantile

71 RAAS: Renin angiotensin-aldosterone system (RAAS)

72 SSA: Sub-Saharan Africa (SSA)

73 TOPMed: Trans-Omics for Precision Medicine

74

75 **1. Introduction**

76 Cardiovascular disease is an exploding epidemic facing low-middle-income countries (LMICs)
77 in Africa, and hypertension is the leading cause of death globally, with the greatest burden
78 of disease in LMICs [1]. ACEI are a class of drugs that inhibit the renin-angiotensin-
79 aldosterone system (RAAS), with a proven reduction in mortality from hypertension,
80 diabetes mellitus, and cardiac failure [2]. They are widely available and affordable, making
81 them critical for use in LMICs [3]. ACEI use is limited by two major adverse events: ACEI-
82 angioedema (AE-ACEI) and cough (ACEI-cough) [4]. AE-ACEI typically involves the face,
83 tongue, or larynx and can be life-threatening in ~16% of cases [5]. AE-ACEI incidence ranges
84 from 0.2-0.7% in retrospective studies to 6% in prospective clinical trials [4-6]; in the only
85 large multicentre African study (CREOLE), the incidence was 0.7% [7]. AE-ACEI is the most
86 common angioedema presentation to emergency rooms across the world, including South
87 Africa [4, 8]. African Americans have a 5-fold increased risk of AE-ACEI compared to
88 European populations [4], and this has led several international hypertension guideline
89 groups to favour angiotensin receptor blockers (ARBs) over ACEI in African populations. We
90 have recently argued that these recommendations are potentially flawed with dire
91 consequences given the near-complete absence of studies of AE-ACEI across diverse African
92 populations [6]. This work aims to address this important gap.

93 Several GWAS and candidate gene studies have been conducted for AE-ACEI [9, 10], with
94 the largest meta-analysis of eight of these cohorts recently published and identifying three
95 SNPs at genome-wide significance (*rs6687813*, *rs35136400*, and *rs6060237* on
96 chromosomes 1, 14, and 21, respectively). Further mapping and gene-based tests suggest
97 that regulatory effects on *the BDKBR2* and *BDKRB1* genes are the most likely underlying
98 mechanisms for the association on chromosome 14. The genes most likely associated with
99 20 SNPs on chromosome 1 are *F5* and *PROCR*, encoding endothelial protein C receptor [11].
100 Other candidate genes with reported associations have been linked with the activities of
101 alternative bradykinin metabolising enzymes (*XPNPEP2* (*rs3788853*), *MME* (*rs989692*),
102 immune regulatory pathways (*PRKCQ* (*rs500766*) and *ETV6* (*rs2724635*)). Few of these
103 polymorphisms have been confirmed functionally or replicated across diverse populations.
104 At present, there have been only two candidate gene association studies (with less than 50
105 AE-ACEI cases and <250 hypertensive patients on ACEI) from Southern Africa that have
106 studied AE-ACEI and ACEI-responsiveness genomics in Sub-Saharan Africa (SSA)[12, 13].
107 These studies associated SNPs *rs1042714* in the *ADRB2* gene, *rs1799722* in the *BDKRB2*
108 gene, and the B₂ receptor -9 allele in the *BDKRB2* gene with AE-ACEI [12, 13]. Furthermore,
109 two lines of evidence support the hypothesis that ACE biology and genomics may vary
110 substantially across the African continent. First, carboxypeptidases, including ACE, are
111 important in shaping the immunopeptidomes of class-I HLA [14], and *Choudhury et al.* [15]
112 found the HLA region to be highly differentiated across African genomic regions. Thus, with
113 epistatic association between HLA and ACE, ACE genomics may vary substantially across SSA
114 populations. Second, polymorphisms in ABO blood group genes have been associated with
115 ACEI-cough, and it is hypothesized that oligosaccharide moieties, acted on by ABO-encoded
116 glycosyltransferases, impact ACE solubility and protease degradation [16]. *ABO* genes have

117 been under substantial selection pressure in Africa due to links with malaria susceptibility
118 [17], and therefore this may be another important mechanism for regional differences in
119 ACE across Africa. This preliminary GWAS from a South African population was aimed at
120 addressing this current research gap.

121

122 **2. Materials and Methods**

123 **2.1 Ethics Statement**

124 Research was carried out in accordance with the latest update of the Declaration of Helsinki.
125 Written informed consent was obtained from all participants in the AE-ACEI and control
126 cohort. The study protocol was approved by the University of Cape Town Faculty of Health
127 Sciences Human Research Ethics Committee (HREC 057/2020) and the Human Research
128 Ethics Committee (Medical) of the University of Witwatersrand (M2210108).

129 **2.2 ACEI-Angioedema Cohort Description**

130 A total of 207 adult patients who had experienced AE-ACEI were recruited retrospectively in
131 the Western (Cape Town area) and Eastern (Mthatha and Gqeberha areas) Cape provinces
132 of South Africa. Patients were defined as having AE-ACEI if they had angioedema while
133 taking an ACEI, with no preceding episodes of angioedema in the absence of ACEI use, and
134 no recurrence of angioedema after removal of the offending ACEI. All cases were reviewed
135 and adjudicated by a clinical expert in allergology. Samples were also obtained from 460
136 controls who had taken an ACEI for at least two years without signs or symptoms of AE.
137 Participants provided saliva collected in an OG-600 kit (DNA Genotek) that was stored at
138 room temperature until DNA extraction. The data collected included demographics and
139 clinical history. All samples were anonymised and labelled with random study identifiers,
140 and the collected data was de-identified for analysis. Seven cases and 92 controls from the

141 Soweto site (Gauteng province) of the Africa Wits-INDEPTH Partnership for Genomic
142 Research (AWI-Gen) study cohort in the Gauteng province were added to this cohort [18,
143 19].

144 **2.3 Genotyping and Quality Control**

145 Participants from the Western/Eastern Cape cohort provided saliva collected in an OG-600
146 kit (DNA Genotek) that was stored at room temperature until DNA extraction. DNA was
147 extracted from saliva using prepIT^oL2P according to manufacturer instructions (DNA
148 Genotek, cat#PT-L2P). All cases and controls were genotyped on the InfiniumTM H3Africa v2
149 array on an Illumina iScan instrument (Illumina, CA, USA) (<https://chipinfo.h3abionet.org>)
150 that has a total of 2,225,121 autosomal SNPs. We used GenomeStudio v2, a software
151 provided by Illumina, to assign genotypes to the raw data and evaluate the Infinium assay
152 controls. H3Africa cluster file v2, generated by Illumina, was used to cluster the genotypes.
153 The PLINK [20] plugin provided by Illumina and supported by GenomeStudio was used to
154 convert the genotype data into PLINK data format to allow for further quality controls using
155 PLINK.

156 A sex discrepancy check was conducted by calculating the X-chromosome homozygosity
157 rate. Any self-identified male that had a homozygosity estimate of $F > 0.8$ and any self-
158 identified female that had an estimate of $F < 0.2$ were removed from the analysis. Genotype
159 and individual missingness in the genotype and samples, respectively, were then checked,
160 and all genotypes and samples that had < 0.02 missingness were removed from the dataset.
161 1,919,455 SNPs and 635 samples were retained after these quality control checks. All our
162 remaining samples were within ± 3 standard deviations from the sample heterozygosity rate
163 mean and were all retained after the heterozygosity check. Hardy-Weinberg Equilibrium
164 (HWE) p-values were then calculated from the controls, and all the SNPs with a p-

165 value $<1.0 \times 10^{-6}$ were further removed, retaining 1,914,061 SNPs. The sample relatedness
166 was then checked from an independent set of SNPs that had been pruned for linkage
167 disequilibrium (LD) by calculating the proportion of identity by descent (IBD) relatedness, π -
168 *hat*, for each pair of samples. The recommended cut-off in GWAS is to use a π -*hat* threshold
169 of 0.1875 to exclude related samples [21]. However, most of the current GWAS methods are
170 designed to control for sample relatedness, particularly those that implement the mixed
171 linear model's association (MLMA) [22]. As GCTA-MLMA [23] was to be used for the GWAS
172 analysis, a lenient threshold was set, and one of the pairs of samples with a π -*hat* > 0.5 was
173 excluded. This retained 616 samples (195 cases and 421 controls). Common SNPs were then
174 extracted from the dataset, and 1,333,573 SNPs that had a minor allele frequency (MAF) $>$
175 0.05 were retained. The AWI-Gen data was subjected to the same QC procedure. The
176 Western/Eastern Cape dataset was then merged with the AWI-Gen data. The final analysis
177 includes 944,944 common SNPs, 202 cases, and 513 controls. Samples were then clustered
178 based on both the self-reported race and ethnicity and also by the phenotype status of the
179 participants by principal component analysis (PCA) using GCTA and plotted using GENESIS
180 [24]. **Figures 1A** and **B** show the PCA plots, respectively, while **Table IS** (*supplementary*
181 *material*) is the demographic table of the study cohort.

182 **2.4 Imputation and GWAS Analysis**

183 The South African cohort was imputed using the African Genome Resource (AGR) panel in
184 the Sanger Imputation Server [25]. This was based on the results of a study of 11,000 sub-
185 Saharan Africans, where $>90\%$ of the samples were genotyped using the H3Africa Array,
186 which showed the AGR and Trans-Omics for Precision Medicine (TOPMed) panels performed
187 best in this population [26]. Further quality control was performed on the imputed dataset,
188 where only the SNPs and samples that had <0.02 missingness on the genotype and samples,

189 respectively, and common SNPs with $MAF > 0.05$ were retained. The imputed dataset after
190 quality control had 7,482,056 SNPs and 715 samples. The imputed data was also clustered
191 based on the self-identified ancestry background, and the PCA plot shown in **Figure 1S**
192 (*supplementary material*) was generated. Standard GWAS was then conducted on the
193 imputed dataset using GCTA-MLMA while controlling for age, sex, global ancestry (using 5
194 principal components (PCs)), and whether or not the samples had hypercholesterolemia,
195 HIV, previous tuberculosis, or asthma, as these were found to be significantly different
196 between the cases and controls (**Table IS**, *supplementary material*). The number of PCs used
197 was determined by obtaining a scree plot for the proportion of variance explained by the
198 first 100 PCs, where 5 PCs were selected using the elbow method (**Figure 2S**, *supplementary*
199 *material*).

200 **2.5 Meta-Analysis**

201 A meta-analysis was performed with the South African and American cohort GWAS
202 summary statistics. The data from the Vanderbilt/Marshfield cohort is publicly available by
203 request on dbGAP under study accession number *phs000438.v1.p1*. It consisted of 546,556
204 autosomal SNPs that had been genotyped using the 610Quadv1.B BeadChip (Illumina, San
205 Diego, California, USA). A build liftover from hg18 to hg19 was first conducted using the
206 liftOver script that is publicly provided by the Centre for Statistical Genetics at the University
207 of Michigan. This cohort was imputed using the Consortium on Asthma among African-
208 ancestry Populations in the Americas (CAAPA) reference panel on the Michigan Imputation
209 Server [27]. The panel consists of 883 African American individuals. After a quality check on
210 the data, where genotypes and samples that had a missingness quality of < 0.02 and rare
211 SNPs with $MAF < 0.05$ were excluded, 5,222,201 SNPs and the 663 samples were retained. A
212 standard GWAS of this cohort was then conducted using GCTA-MLMA with sex and two PCs

213 as covariates, and genomic control (GC) was performed on the p-values. The total sample
214 size for the meta-analysis was thus 376 AE-ACEI cases and 1002 controls. 4,589,885 SNPs
215 that were in common in the South African and Vanderbilt/Marshfield cohorts were selected.
216 METASOFT [28] was first implemented to estimate the heterogeneity of the study, which
217 was found to be 0, and thus a fixed effect model in METASOFT was used for the meta-
218 analysis.

219 **2.6 Functional Annotation and Gene-Based Tests**

220 Functional annotation of the SNPs and prioritization of genes were performed using FUMA
221 (v1.5.2) [29], while MAGMA (Multi-marker Analysis of GenoMic Annotation), implemented
222 in FUMA, was used in the gene-based tests. All the SNPs from the GWAS of the South
223 African cohort and the meta-analysis were annotated. The 1000 genomes Phase3 African
224 population reference panel LD backgrounds were used, and a p-value threshold of 1.0×10^{-05}
225 was set for the lead SNPs in FUMA.

226 **2.7 Replication**

227 A number of AE-ACEI association studies on angioedema have been conducted to date [9,
228 30-33]. In particular, *Liau et al.* [30] have identified more than 10 GWAS for AE-ACEI.
229 Cumulatively, these studies have highlighted a total of 75 SNPs that have been found to be
230 associated with angioedema. We sought to replicate some of the SNPs detected in both the
231 standard GWAS and the meta-analysis.

232

233 **3. Results**

234 The AE-ACEI cases (n=202) and controls (n=513) used in the standard GWAS and meta-
235 analysis were similar in terms of age (55.5 ± 17.8 years vs. 59 ± 18.0 years, respectively, p-
236 value=0.16), predominantly female sex (63% [128/202] vs. 52% [269/513] respectively, p-

237 value=0.008), and co-morbid illness. The AE-ACEI cases had significantly more patients who
238 self-reported as black at 58.6% (116/202), than the ACEI-tolerant controls at 43.3%
239 (224/513) (p-value<0.001) (**Table IS, supplementary material**).

240 **3.1 GWAS Association and Functional Annotation of the South African Cohort**

241 The Manhattan and QQ plots of the post-imputation South African GWAS are shown in
242 **Figures 2A and B**, respectively. In this analysis, 26 SNPs (**Table I**) were detected at a
243 suggestive threshold (p-values<5.0x10⁻⁰⁶). Additionally, no inflated p-values were observed
244 in the post-imputation GWAS analysis ($\lambda=0.99$). As is common in GWAS, most of the 26 SNPs
245 detected at a suggestive threshold were found to be either intergenic or intronic variants
246 and included both imputed and genotyped variants (**Table I**). These were also found to be
247 close to 14 genes by positional mapping. The CUB and Sushi multiple domains 1 gene,
248 *CSMD1*, reported by *Hallberg et al.* [34] in the GWAS catalog, has been linked to ACEI-
249 induced cough. None of the other variants have previously been linked to angioedema.

250 **3.2 Meta-analysis and Functional Annotation**

251 The Manhattan and QQ plots of the standard GWAS of the Vanderbilt/Marshfield cohort
252 post-imputation are shown in **Figures 3S A and B (supplementary material)**. To correct for
253 the effect of population structure, the p-values were GC-corrected, and the inflation factor
254 improved from $\lambda=0.56$ to $\lambda_{GC}=1.0$. Similar to a previous GWAS on this cohort by *Pare et al.*
255 [35], none of the SNPs studied were observed to be significant; however, 25 SNPs with a p-
256 value<5.0x10⁻⁰⁶ were detected. Using a pre-imputed dataset of this cohort, *Pare et al.* [35]
257 performed the association study using conditional logistic regression stratified by ancestry
258 and obtained 41 SNPs that had a p-value<1.0x10⁻⁰⁴. They highlighted two of the SNPs,
259 *rs500766* on chromosome 10 and *rs2724635* on chromosome 12, which they linked to the
260 *PRKCQ* and *ETV6* genes, respectively. We observed *rs500766* at p-value=2.14x10⁻⁰⁵ and

261 *rs2724635* at $p\text{-value}=1.71\times 10^{-03}$ in our post-imputation standard GWAS analysis of the
262 cohort. The fixed-effect meta-analysis of the South African and Vanderbilt/Marshfield
263 cohorts resulted in the Manhattan and QQ plots in **Figures 3A** and **B**. No p -value inflation of
264 p -values was observed in the analysis ($\lambda=0.99$). The meta-analysis detected 37 SNPs with a
265 $p\text{-value}<5.0\times 10^{-06}$ and were observed to be mainly intergenic or intronic by positional
266 mapping; see **Table II**.

267 In addition to detecting SNP *rs7815832* ($p\text{-value}=8.26\times 10^{-07}$) located near gene *CSMD1* as
268 was observed in the standard GWAS, the meta-analysis detected SNP *rs4750617* (p -
269 $\text{value}=2.92\times 10^{-06}$) about 56 kb upstream of SNP *rs500766* located in the *PRKCQ* (protein
270 kinase C theta) gene on chromosome 10 that was linked to AE-ACEI by *Pare et al* [35]. The
271 meta-analysis also detected 9 SNPs at a suggestive threshold on chromosome 6 that were
272 located close to *RIMS1*, the regulating synaptic membrane exocytosis 1 gene, which has
273 previously been linked to angioedema in a Spanish population [36].

274 **3.3 Replication Analysis**

275 In this analysis, a Bonferroni-corrected significance threshold of 6.67×10^{-04} ($0.05/75$) was
276 used. Though SNP *rs34485356* previously mapped to *BDKRB2* was observed in the standard
277 GWAS with a p -value of 0.004 and SNP *rs500766* located in gene *PRKCQ* was observed at a
278 p -value of 0.009 in the meta-analysis, none of the SNPs in both studies were replicated at a
279 significant threshold.

280 **3.4 Gene-based Test**

281 In total, all the input SNPs in FUMA for the standard GWAS and meta-analysis were mapped
282 to 18,853 and 18,093 protein-coding genes, respectively. The gene-based test using
283 MAGMA for the standard GWAS thus considered a Bonferroni-corrected significance
284 threshold of 2.65×10^{-06} and 2.76×10^{-06} for the two tests, respectively. None of the genes

285 considered were found to be significant. **Tables IIS** and **IIIS** (*supplementary material*) list the
286 top 10 genes highlighted by MAGMA in each analysis, respectively, while **Figures 4S** and **5S**
287 (*supplementary material*) are the corresponding Manhattan plots.

288

289 **4. Discussion**

290 Our results highlight the importance of African populations to detect novel variants and
291 potentially replicate preliminary signals from other populations. To our knowledge, this is
292 the largest GWAS to investigate AE-ACEI in a diverse African cohort on the continent. The
293 underrepresentation of continental African populations in GWAS is concerning [37]. This
294 may lead to healthcare disparities once GWAS results are translated into clinical relevance,
295 as well as limit our understanding of the still-missing heritability that continues to plague
296 GWAS and impact the accuracy of predicting drug responses in diverse populations [38]. In
297 the context of ACEI use, we have highlighted the influence that early epidemiological
298 evidence from African American studies has had on international hypertension guidelines
299 and have warned against the potential pitfalls of extrapolating this very limited data to all
300 continental African populations [6].

301 In this study, we have performed both a standard GWAS analysis in the South African
302 population and meta-analysed our study with an African American and European American
303 cohort from Vanderbilt (Nashville, Tennessee) and Marshfield (Wisconsin). We have further
304 implemented FUMA to annotate all the SNPs and conducted a gene-based test for the
305 protein-coding genes that were found close to these SNPs. Our standard GWAS of the South
306 African cohort and the meta-analysis detected 26 SNPs and 37 SNPs at suggestive thresholds
307 ($p\text{-value} < 5.0 \times 10^{-6}$), respectively, which were located close to 23 genes by positional
308 mapping in FUMA. Among them, the *RIMS1* gene has been associated with non-steroidal

309 anti-inflammatory drug (NSAID)-induced angioedema [36]. *RIMS1* modulates G-proteins, in
310 particular those linked to the opening of calcium channels, and this has been best studied in
311 relation to the release of neurotransmitters and insulin [39]. *BDKRB1* and *BDKRB2* are both
312 G-protein coupled receptors, and therefore *RIMS1* may play a role in modulating bradykinin
313 receptor-2 sensitivity in susceptible individuals [40]. In addition, in a GWAS study of smoking
314 patterns and meta-analysis of smoking status, *Xu et al.* [41] found a significant variant
315 *rs1334346* ($p\text{-value}=8.22\times 10^{-09}$) close to *RIMS1* that was associated with smoking behaviour
316 over time. The *CSMD1* gene, which is a complement regulatory protein linked to kallikrein in
317 pathway analyses, was also found close to the SNPs detected at a suggestive threshold and
318 has been reported by *Hallberg et al.* [34] and *Saunders et al.* [42] on the GWAS catalog.
319 *CSMD1* is linked to cough in response to ACEI drugs and to age at initiation of smoking,
320 respectively. Smoking has been epidemiologically identified as a risk factor for AE-ACEI [50],
321 but we did not capture smoking status in our clinical data. The *PRKCQ* gene on chromosome
322 10, which is associated with T-cell activation [43], was a signal highlighted in the original
323 Vanderbilt/Marshfield cohort analysis [35] to be linked to AE-ACEI. As noted in the recent
324 meta-analysis of AE-ACEI in European participants, there is now an urgent need for
325 functional data to confirm the biological role of some of these associated genes, particularly
326 those impacting bradykinin receptor sensitivity and signalling channels.

327 The main limitation of our study is the small sample size and, thus, the low power to attain
328 genome-wide significance for novel SNP associations. However, these findings highlight the
329 importance of the inclusion of African GWAS in replication and meta-analysis. The large
330 number of SNPs obtained at a suggestive threshold in both the post-imputation standard
331 GWAS and the meta-analysis further highlights the need for increased sampling on the
332 continent if African GWAS is to catch up with European GWAS.

333 In conclusion, this study presents the largest GWAS of AE-ACEI from a continental African
334 population, with several SNPs detected at a suggestive threshold in both the post-
335 imputation GWAS and meta-analysis, including SNPs near genes with biological plausibility
336 and prior associations with drug-induced angioedema. Further work is now required to
337 increase sampling across diverse African regions to improve study power and further
338 illuminate the heritability of AE-ACEI.

339

340 **Perspectives**

341 We performed a GWAS of a South African cohort and a meta-analysis with summary
342 statistics of an African American and European American cohort. We located SNPs
343 associated with AE-ACEI at a suggestive threshold ($p\text{-value} < 5.0 \times 10^{-06}$) close to the *CSMD1*
344 gene, previously linked to ACEI cough, as well as the *RIMS1* and *PRKCQ* genes linked to drug-
345 induced angioedema. The study highlights the importance of African populations in meta-
346 analysis and replication studies and the need for increased sampling on the continent of
347 Africa.

348

349 **Novelty and Relevance**

350 **What is new?**

351 To our knowledge, this is the largest GWAS and meta-analysis study on AE-ACEI in a South
352 African cohort that has replicated genes linked to drug-induced angioedema.

353 **What is Relevant?**

354 The study identifies SNPs with suggestive significance thresholds that are associated with
355 AE-ACEI. It also highlights the need for increased sampling on the African continent to
356 ensure novel results and thus equitable healthcare when GWAS hits are translated to clinical
357 relevance.

358 **Clinical/Pathophysiological Implications?**

359 Identification of risk variants associated with AE-ACEI in African populations will better
360 inform hypertension treatment guidelines for this population.

361

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373 **Disclosures**

374 None.

375 **Supplementary data**

376 Supplementary Figures (**1S-5S**) and Tables (**IS-IIIS**) are available.

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Table I: List of 26 SNPs that were obtained at a suggestive threshold ($p\text{-value} < 5.0 \times 10^{-06}$) in the post-imputation standard GWAS, the corresponding list of genes they were found close to the SNPs by positional mapping, their function, and whether the SNP was genotyped or imputed.

CHR	rsID	Position	A1	A2	MAF	Pvalue	Nearest Gene	Function	Genotyped/Imputed
1	rs6666273	191428753	A	G	0.23	4.58×10^{-06}	<i>RP11-309H21.2</i>	intergenic	Imputed
1	rs4660011	241039771	C	A	0.44	1.72×10^{-06}	<i>RGS7</i>	intronic	Imputed
1	rs12239109	241042754	C	T	0.37	4.59×10^{-06}	<i>RGS7</i>	intronic	Genotyped
1	rs10926374	241047316	C	T	0.38	4.94×10^{-06}	<i>RGS7</i>	intronic	Imputed
1	rs7544945	241048562	T	C	0.38	4.92×10^{-06}	<i>RGS7</i>	intronic	Imputed
1	rs12127148	241049298	G	T	0.37	3.9×10^{-06}	<i>RGS7</i>	intronic	Imputed
5	rs28630061	176204086	T	C	0.16	1.27×10^{-07}	<i>RP11-375B1.3</i>	intergenic	Imputed
8	rs62488207	4808991	T	G	0.34	4.79×10^{-06}	<i>CSMD1</i>	intronic	Imputed
8	rs12544185	4812064	A	T	0.33	3.32×10^{-06}	<i>CSMD1</i>	intronic	Imputed
8	rs7815832	4812436	A	C	0.35	1.24×10^{-07}	<i>CSMD1</i>	intronic	Genotyped
9	rs142454594	10278823	A	G	0.07	1.50×10^{-06}	<i>PTPRD</i>	intronic	Imputed
9	rs79552273	10279110	T	C	0.07	4.91×10^{-06}	<i>PTPRD</i>	intronic	Genotyped
11	rs12287265	60237028	A	T	0.20	2.41×10^{-07}	<i>MS4A1</i>	UTR3	Imputed

11	<i>rs75124932</i>	60250467	C	G	0.20	6.28×10 ⁻⁰⁷	<i>MS4A12</i>	intergenic	Imputed
12	<i>rs73409491</i>	124545501	A	G	0.13	3.22×10 ⁻⁰⁶	<i>FAM101A</i>	intronic	Genotyped
12	<i>rs58371925</i>	124550518	C	T	0.14	2.69×10 ⁻⁰⁶	<i>FAM101A</i>	intronic	Imputed
12	<i>rs58449918</i>	124551080	G	T	0.13	2.69×10 ⁻⁰⁶	<i>FAM101A</i>	intronic	Imputed
12	<i>rs112593781</i>	124559622	G	A	0.14	1.34×10 ⁻⁰⁶	<i>FAM101A</i>	intronic	Imputed
13	<i>rs116365021</i>	34778472	G	A	0.06	4.19×10 ⁻⁰⁶	<i>SNORA25</i>	intergenic	Imputed
13	<i>rs28810452</i>	38841040	A	G	0.19	3.53×10 ⁻⁰⁶	<i>UFM1</i>	intergenic	Imputed
13	<i>rs58482084</i>	38845507	G	A	0.19	3.53×10 ⁻⁰⁶	<i>UFM1</i>	intergenic	Imputed
13	<i>rs59296363</i>	112101252	T	G	0.12	6.00×10 ⁻⁰⁷	<i>TEX29</i>	intergenic	Genotyped
14	<i>rs2145156</i>	69095279	A	G	0.08	3.39×10 ⁻⁰⁶	<i>RAD51B</i>	intronic	Imputed
17	<i>rs73321625</i>	45977502	T	A	0.16	1.05×10 ⁻⁰⁶	<i>SP2:AC003665.1</i>	ncRNA exonic	Imputed
17	<i>rs73321628</i>	45977632	G	A	0.16	1.05×10 ⁻⁰⁶	<i>SP2:AC003665.1</i>	ncRNA exonic	Genotyped
19	<i>rs12610494</i>	8407470	C	G	0.09	1.07×10 ⁻⁰⁶	<i>KANK3</i>	intronic	Imputed

***CHR**-chromosome, ***A1**-minor allele, ***A2**-reference alleles, ***MAF**-minor allele frequency

Table II: The list of 37 SNPs that were obtained at a suggestive threshold ($p\text{-value} < 5.0 \times 10^{-6}$) in the meta-analysis, the corresponding list of genes they were found close to by positional mapping, their function, and whether the SNP was genotyped or imputed.

CHR	rsID	Position	A1	A2	MAF	P-value	Nearest Gene	Function	Genotyped/Imputed
2	rs6726307	72030471	G	A	0.2	4.51×10^{-06}	<i>DYSF</i>	intergenic	Genotyped
5	rs3924097	120445067	T	A	0.4	3.6×10^{-06}	<i>CTD-261308.1</i>	intergenic	Imputed
6	rs34388291	40580161	A	G	0.2	4.31×10^{-06}	<i>LRFN2</i>	intergenic	Imputed
6	rs34207487	40582609	T	C	0.21	7.23×10^{-07}	<i>LRFN2</i>	intergenic	Imputed
6	rs1192177	72572886	T	C	0.5	3.01×10^{-06}	<i>RIMS1</i>	intergenic	Imputed
6	rs1192178	72573194	A	G	0.5	3.01×10^{-06}	<i>RIMS1</i>	intergenic	Imputed
6	rs1192179	72573244	C	T	0.5	3.01×10^{-06}	<i>RIMS1</i>	intergenic	Imputed
6	rs1763305	72576793	G	T	0.5	4.28×10^{-06}	<i>RIMS1</i>	intergenic	Imputed
6	rs1612461	72576794	A	C	0.5	4.28×10^{-06}	<i>RIMS1</i>	intergenic	Imputed
6	rs1147527	72579416	A	G	0.49	3.59×10^{-06}	<i>RIMS1</i>	intergenic	Imputed
6	rs1147526	72579596	G	C	0.5	4.28×10^{-06}	<i>RIMS1</i>	intergenic	Imputed
6	rs1192182	72581129	G	A	0.5	4.28×10^{-06}	<i>RIMS1</i>	intergenic	Imputed
6	rs1147522	72583600	T	C	0.47	1.28×10^{-06}	<i>RIMS1</i>	intergenic	Genotyped
7	rs6952303	119603778	T	C	0.46	1.35×10^{-06}	<i>U1</i>	intergenic	Imputed

7	rs6953016	119604064	A	G	0.46	1.35×10^{-06}	U1	intergenic	Imputed
7	rs10281213	119605536	T	C	0.46	1.74×10^{-06}	U1	intergenic	Genotyped
7	rs7802501	119629586	G	A	0.46	1.35×10^{-06}	U1	intergenic	Genotyped
7	rs10236345	119631602	C	A	0.46	1.35×10^{-06}	U1	intergenic	Imputed
7	rs10257336	119635285	A	G	0.46	1.62×10^{-06}	U1	intergenic	Imputed
7	rs10247779	119639646	G	A	0.46	1.62×10^{-06}	U1	intergenic	Imputed
7	rs58094128	119651727	C	T	0.46	2.62×10^{-06}	U1	intergenic	Imputed
7	rs10245443	119656806	T	G	0.47	7.70×10^{-07}	U1	intergenic	Imputed
8	rs7815832	4812436	C	A	0.35	8.26×10^{-07}	CSMD1	intronic	Imputed
10	rs4750617	6606334	A	C	0.36	2.92×10^{-06}	PRKCQ	intronic	Imputed
11	rs12287265	60237028	A	T	0.2	1.70×10^{-07}	MS4A1	UTR3	Imputed
18	rs4553714	44298996	A	G	0.16	3.54×10^{-06}	ST8SIA5	intronic	Imputed
20	rs6021761	50776522	C	A	0.34	4.41×10^{-06}	ZFP64	intronic	Imputed
21	rs2826127	21666445	G	A	0.38	3.45×10^{-06}	AP001171.1	intergenic	Imputed
21	rs9981281	21666624	T	C	0.38	3.45×10^{-06}	AP001171.1	intergenic	Imputed
21	rs1077650	21666756	A	G	0.38	4.27×10^{-06}	AP001171.1	intergenic	Imputed
21	rs1077651	21666997	C	T	0.38	3.89×10^{-06}	AP001171.1	intergenic	Genotyped

21	rs1027063	21667752	G	C	0.38	3.89×10^{-06}	AP001171.1	intergenic	Imputed
21	rs1027064	21667841	T	A	0.38	3.89×10^{-06}	AP001171.1	intergenic	Imputed
21	rs1027065	21667851	C	T	0.38	4.62×10^{-06}	AP001171.1	intergenic	Imputed
21	rs1027066	21667880	A	G	0.38	3.89×10^{-06}	AP001171.1	intergenic	Imputed
21	rs1028981	21668483	A	G	0.38	4.27×10^{-06}	AP001171.1	intergenic	Imputed
21	rs1028982	21668534	A	G	0.38	4.09×10^{-06}	AP001171.1	intergenic	Imputed

***CHR**-chromosome, ***A1**-minor allele, ***A2**-reference alleles, ***MAF**-minor allele frequency.

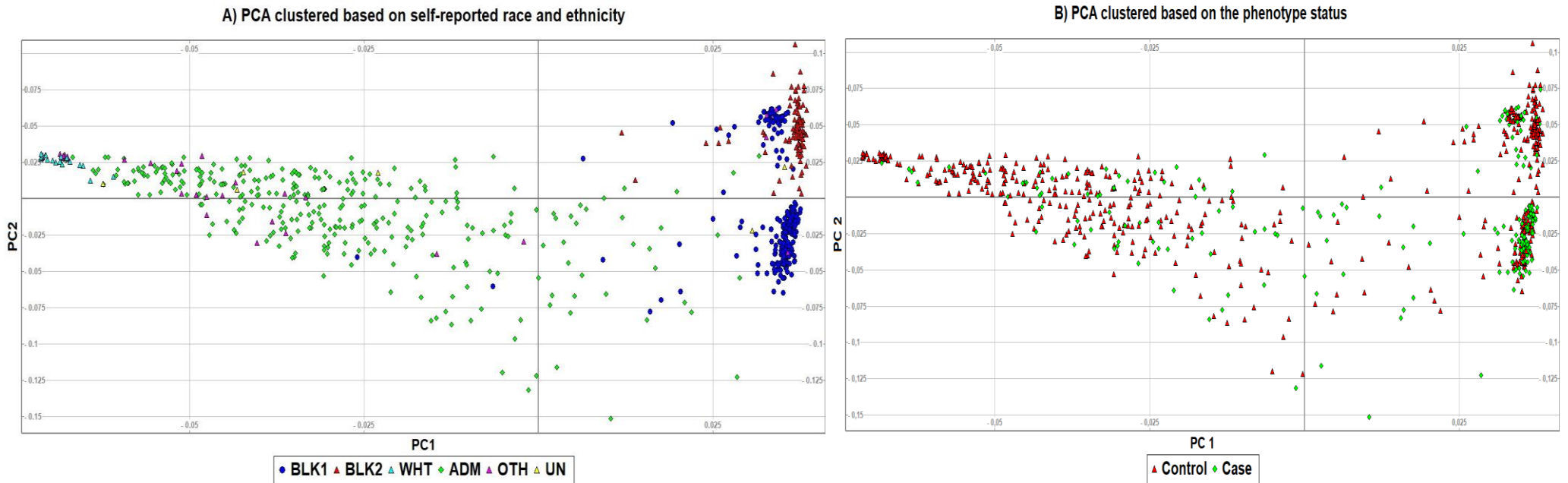


Figure 1: The PCA plot of the South African cohort clustered based on **A)** self-identified race and ethnicity: BLK1 – Black population from Western and Eastern Cape, BLK2 – Black population from Soweto, WHT – White, ADM – mixed ancestry, OTH- Other, UN – Unreported, and **B)** AE-ACEI status. This highlights the genomic diversity and the distribution of the cases and controls in study cohort.

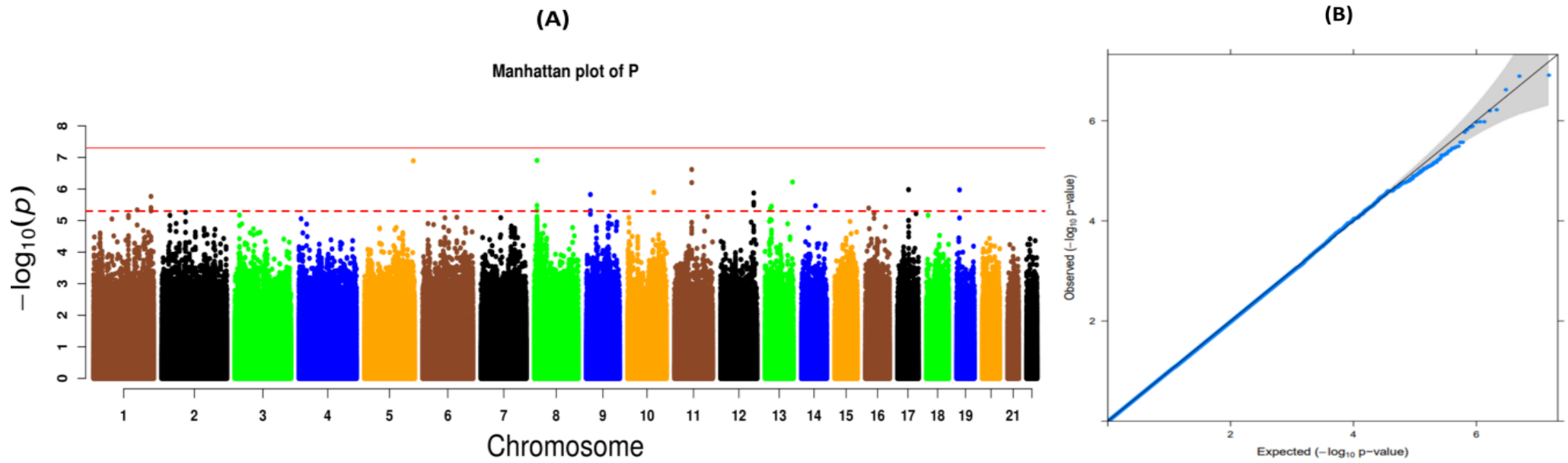


Figure 2: A) The Manhattan plot of the standard GWAS of the South African cohort. The dashed red line corresponds to $p\text{-value}=5.0 \times 10^{-6}$, while the solid line is the GWAS significance threshold $p\text{-value}=5.0 \times 10^{-8}$. **B)** The corresponding QQ plot of the p -values of the standard GWAS ($\lambda=0.99$).

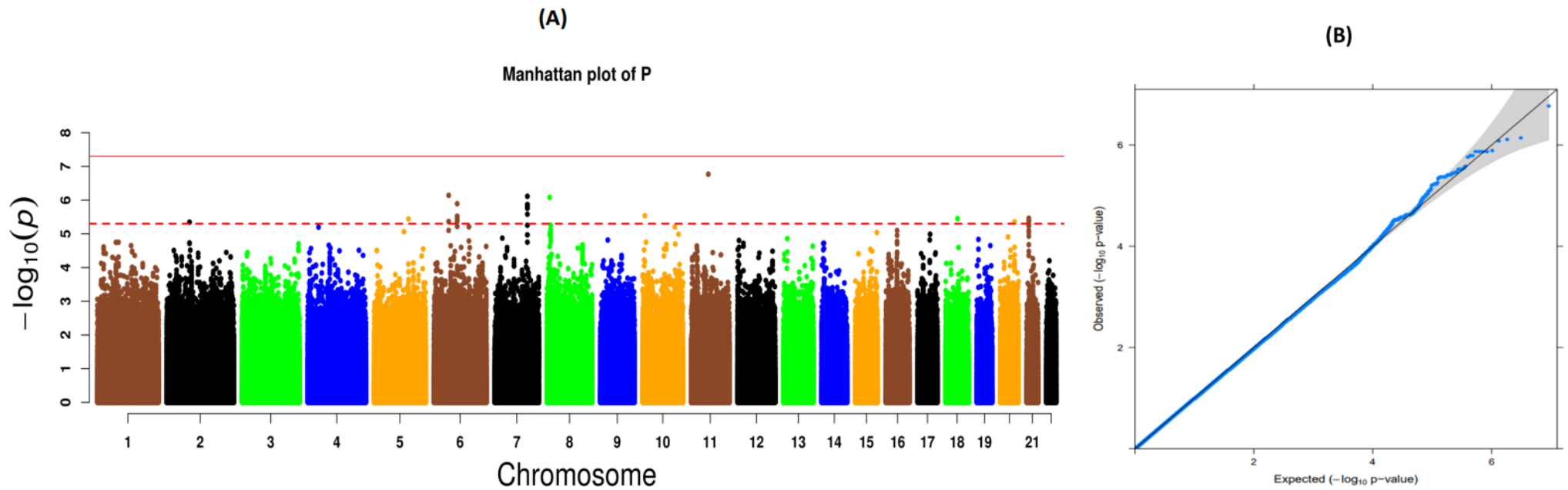


Figure 3: A) The Manhattan plot of the meta-analysis of the South African and Vanderbilt/Marshfield cohort summary statistics. The dashed red line corresponds to $p\text{-value}=5.0 \times 10^{-6}$, while the solid line is the GWAS significance threshold $p\text{-value}=5.0 \times 10^{-8}$. **B)** The corresponding QQ plot of the p-values of the standard GWAS ($\lambda=0.99$).