1 The association of the *ACTN3* R577X and *ACE* I/D polymorphisms with 2 athlete status in football: a systematic review and meta-analysis

3 Alexander B. T. McAuley¹, David C. Hughes¹, Loukia G. Tsaprouni¹, Ian

4 Varley², Bruce Suraci³, Thomas R. Roos⁴, Adam J. Herbert¹, and Adam L. Kelly¹

5 ¹Faculty of Health, Education and Life Sciences, Birmingham City University, Birmingham,

6 West Midlands, United Kingdom; ²Department of Sport Science, Nottingham Trent University,

7 Nottingham, United Kingdom; ³Academy Coaching Department, AFC Bournemouth,

8 Bournemouth, United Kingdom; ⁴The International Academy of Sports Science and

9 Technology (AISTS), University of Lausanne, Lausanne, Switzerland

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11 Correspondence: A. McAuley, Department of Life Sciences, Birmingham City University,

12 City South Campus, Westbourne Road, Edgbaston, B15 3TN, UK. E-mail:

13 <u>Alex.Mcauley@mail.bcu.ac.uk</u>

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17 The aim of this review was to assess the association of ACTN3 R577X and ACE I/D 18 polymorphisms with athlete status in football and determine which allele and/or genotypes 19 are most likely to influence this phenotype via a meta-analysis. A comprehensive search 20 identified 17 ACTN3 and 19 ACE studies. Significant associations were shown between 21 presence of the ACTN3 R allele and professional footballer status (OR = 1.35, 95% CI: 22 1.18-1.53) and the ACE D allele and youth footballers (OR = 1.18, 95% CI: 1.01-1.38) 23 compared to a control group. More specifically, the ACTN3 RR genotype (OR = 1.48, 95%24 CI: 1.23-1.77) and ACE DD genotype (OR = 1.29, 95% CI: 1.02-1.63) exhibited the 25 strongest associations, respectively. These findings may be explained by the association of 26 the ACTN3 RR genotype and ACE DD genotype with power-orientated phenotypes and the 27 relative contribution of power-orientated phenotypes to success in football. As such, the 28 results of this review provide further evidence that individual genetic variation may 29 contribute towards athlete status and can differentiate athletes of different competitive 30 playing statuses in a homogenous team-sport cohort. Moreover, the ACTN3 R577X and 31 ACE I/D polymorphisms are likely (albeit relatively minor) contributing factors that 32 influence athlete status in football.

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34 Keywords: Soccer; Team-Sport; Genetics; SNP; Genomics.

35 1 Introduction

36 Actinin alpha 3 (ACTN3), a member of the actin family, is a sarcomeric protein which is greatly expressed in muscle tissue ¹. A function of the ACTN3 protein involves crosslinking fast-twitch (type 37 38 II) actin filaments in skeletal muscle fibres². Thus, the expression of the ACTN3 protein in glycolytic 39 skeletal muscle is thought to be a contributing factor to the generation of powerful and explosive muscle 40 contractions; through optimal coordination of type II muscle fibres ³. The coding of the ACTN3 protein 41 is controlled by the ACTN3 gene, located on chromosome 11q13.2. A common genetic variant in the 42 ACTN3 gene has been identified which significantly alters the production of the ACTN3 protein 4 . The 43 genetic variation is a nonsense single nucleotide polymorphism (SNP) which can introduce a premature 44 stop codon within the gene at position 577 (rs1815739)⁵. Cytosine is the most common nucleotide at 45 this position (i.e., $\underline{C}GA$), which encodes the amino acid, arginine (R)². Alternatively, thymine can be 46 possessed by an individual (i.e., TGA), producing the stop codon (X); potentially resulting in an individual being deficient in ACTN3⁴. As the ACTN3 gene portrays a role in force production, it has 47 48 been hypothesised that the performance of activities requiring extensive force production (i.e., 49 sprinting, jumping, weightlifting) would be influenced by whether an individual possesses the R allele 50 or RR genotype. Many studies have reported that either the RR genotype was over-represented, or the 51 XX genotype was underrepresented, in power-related sports (e.g., 100m sprint, rowing, speed skating, 52 artistic gymnastics, sprint swimming, Olympic weightlifting) across American, Polish, Finnish, Italian, Japanese, Israeli, and Russian cohorts ⁶⁻¹⁵. Indeed, Ma and colleagues ¹⁶ conducted a meta-analysis on 53 54 23 studies involving power and endurance athletes and discovered that the R allele was only associated 55 with power athletes. Moreover, a recent meta-analysis on solely power athletes reported similar 56 associations between the R allele and power athletes across 38 studies ¹⁷.

57 Another commonly investigated gene in sport performance is the angiotensin I converting 58 enzyme (ACE) gene. The angiotensin I converting enzyme catalyses the degradation of the inactive 59 decapeptide angiotensin I, and subsequently generates the physiologically active peptide, angiotensin 60 II; an oligopeptide of eight amino acids that binds to specific receptors in the body affecting several systems ^{18,19}. Angiotensin II can constrict blood vessels and stimulate aldosterone production, resulting 61 62 in increased blood pressure, thirst, or the dire for salt. As such, the ACE enzyme is the most crucial 63 component of the renin-angiotensin system (RAS), as it is a potent vasopressor and aldosterone-64 stimulating peptide which regulates blood pressure and fluid-electrolyte balance ²⁰. A polymorphism 65 has been identified within intron 16 of the ACE gene, located on chromosome 17q23.3 66 (NC 000017.11), which results in a substantial variation of RAS activity ^{21,22}. The polymorphism is 67 known as an insertion/deletion (indel) polymorphism, with the insertion (I allele) and deletion (D allele) 68 representing the presence and absence of a 287-bp Alu-sequence respectively. Specifically, the I allele 69 has been associated with lower serum and tissue ACE activity, alongside an increased percentage of 70 slow-twitch (type I) muscle fibres; whilst the D allele has been associated with higher circulating and 71 tissue ACE activity, alongside greater strength and muscle volume and an increased percentage of type II muscle fibres ^{21–23}. In the context of sport, the I allele has been frequently associated with elite 72 73 endurance performance. Specifically, higher I allele frequencies have been reported in middle- and 74 long-distance rowers, swimmers, road-cyclists, runners, mountaineers, cross-country skiers, and tri-75 athletes across a range of diverse cohorts (e.g., British, Australian, Croatian, Russian, Spanish, Italian, Turkish, Polish, Japanese, Indian)^{24–33}. Indeed, during the meta-analysis of Ma and colleagues¹⁶, the 76 77 authors also assessed the influence of the ACE I/D polymorphism on endurance athletes over 25 studies, 78 reporting that the II genotype was significantly associated with endurance athletes. However, the 79 authors found no association between the ACE I/D polymorphism and power athletes. This may have been due to the large heterogeneity observed between studies ($I^2 = >75\%$), most likely a result of not 80 81 analysing the power athletes independently based upon ethnicity. Indeed, a more recent meta-analysis 82 did conduct an ethnic specific analysis of power athletes and reported significant associations with the 83 ACE D allele 34 .

The collective results on the associations of the ACTN3 R577X and ACE I/D polymorphisms, 84 85 with athletic performance, complicate the implications for sports which require both power and 86 endurance related traits, such as football. Football is an intermittent sport which requires optimal utilisation of both the aerobic and anaerobic systems ^{35,36,37,38}; and as such, predicting whether a power 87 88 or endurance-orientated allelic variant may be preferable is not straightforward. Therefore, genetic 89 association studies began to investigate whether a power or endurance-orientated genotype was more 90 important in football by analysing polymorphisms such as ACTN3 R577X and ACE I/D³⁹⁻⁴¹. Moreover, 91 studies began to assess if there was a difference between football players of various competitive playing 92 levels (i.e., elite, non-elite; professional [PRO], non-professional [NP]) and controls (CON), in order to determine if ACTN3 R577X or ACE I/D were associated with athlete status ^{42–44}. Currently, there is no 93 94 general consensus on the importance of ACTN3 or ACE in footballers, with studies reporting positive, 95 negative, and contrasting allelic associations ³⁹⁻⁴⁴. This is most likely because each gene or genotype 96 has a small contribution to sporting performance and is dependent on numerous inter-individual 97 variations (e.g., ethnicity and competitive playing level) ^{45,46}. As a result, studies require large homogenous sample sizes in order to have sufficient statistical power and demonstrate significant 98 99 associations and replications ^{45,47}. However, studies within football genomics have notoriously small 100 sample sizes and many are heterogenic multi-sport studies; mainly due to the unique population and 101 limited access available ⁴⁸. Therefore, to overcome the limitation of sample sizes and heterogeneity, a 102 meta-analysis can be used to pool the results of single homogenous studies together ⁴⁷. As such, the aim 103 of this study was to assess if the ACTN3 R577X and/or ACE I/D polymorphism are associated with 104 athlete status in football by conducting a systematic review and meta-analysis.

105 **2 Methodology**

106 2.1 Search Strategy and Inclusion/Exclusion Criteria

107 In accordance to the Preferred Reporting Items for Systematic reviews and Meta-Analysis (PRISMA) guidelines ⁴⁹, the following search strategy was implemented. A comprehensive search of the Pubmed, 108 SPORTDiscus, and MEDLINE databases was conducted on March 3rd 2020. For ACTN3 the following 109 110 Boolean search was used: ((football) OR (soccer)) AND (actn3) OR (alpha-actinin-3) OR (actinin-111 alpha-3) OR (R577X) OR (rs1815739)). For ACE the following Boolean search was used: ((football) 112 OR (soccer)) AND (ace) OR (angiotensin I converting enzyme) OR (rs1799752) OR (rs13447447) OR 113 (rs4341) OR (rs4646994)). Additionally, Google Scholar was searched using word combinations of the 114 aforementioned Boolean searches, with no year restriction placed on any search. Furthermore, reference 115 lists of the identified articles were searched for additional relevant studies. At the initial screening stage 116 studies were included if they: (1) were primary cohort or case-control investigations; (2) presented 117 ACTN3 R577X or ACE I/D genotype frequencies of footballers in isolation; and (3) were published in 118 the English language. Therefore, studies were excluded if they: (1) were reviews; (2) presented ACTN3 119 R577X or ACE I/D genotype frequencies of footballers combined with other sports; and (3) were 120 published in a language other than in English.

121 **2.2 Data Extraction and Analysis**

122 We extracted the following data from all studies: first author's name and year of publication; number 123 of footballers and CON; nationality and ethnicity; gender; age range; competitive playing level; type of 124 study (cohort or case-control); and distribution of genotype frequencies in footballers and CON. 125 Extracted data was then analysed in the following order: 1) pooled genotype frequencies of case-control 126 studies in isolation; 2) pooled genotype frequencies of case-control studies combined with cohort 127 studies (with an ethnically matched independent CON population added); and 3) sub-group analysis of 128 ethnicity, gender, and level of competition. Independent CONs were added to cohort studies from the 129 1000 genomes database (https://www.internationalgenome.org) or an independent non-included study. 130 Each cohort study was assigned a CON which was not used by any other study in the analysis to bolster 131 the number of unique individuals and decrease selection bias. All of the included CON were also

132 subjected to risk of bias before being included.

133 **2.3 Risk of Bias**

After initial primary inclusion, all studies were subjected to Hardy–Weinberg equilibrium (HWE) via chi-square (significance level P<0.05); culminating in the removal of all studies violating this significance threshold, as deviations from HWE in a CON group can indicate potential genotyping errors, selection bias and stratification ⁵⁰. Furthermore, as each study is tested for HWE, Benjamini-Hochberg false discovery rate (FDR) is used to correct p-values ⁵¹. Additionally, as this review only

139 uses published studies, publication bias could be a limiting factor. Therefore, Egger's test ⁵² in

140 combination with funnel plots 53 were used to identify if publication bias was present and potentially 141 skewing results (significance level P<0.05).

142 2.4 Statistical Analysis

To measure the strength of the association between the ACTN3 R577X and ACE I/D polymorphisms 143 144 and football players, odds ratios (OR), 95% confidence intervals (CI), and forest plots were used via 145 either a fixed-effects model or random-effects model to identify the individual and pooled effects of the 146 studies. Determining which model was appropriate for each analysis was based on the level of heterogeneity revealed via the l^2 (<50% = fixed-effects model; >50% = random-effects model) and 147 Cochran's Q⁵⁴ statistical test (significance level at P<0.05). Four genetic models were used to assess 148 149 genotype and allele differences between football players and CONs: (a) allele contrast; (b) recessive; 150 (c) dominant; and, (d) over-dominant. Additionally, pair-wise comparisons were also conducted. 151 Finally, a sensitivity analysis was conducted via a leave-one-out approach in order to test the robustness of results ⁵⁵. All analyses were conducted using the MetaGenyo online Statistical Analysis System 152 153 software (http://bioinfo.genyo.es/metagenyo/) 56.

154 **3 Results**

155 **3.1** Search Process

156 The systematic search processes initially identified 588 studies (*ACTN3*, n = 290; *ACE*, n = 298). 157 Following the removal of duplicates and the screening of titles and abstracts, full-text assessment 158 commenced; culminating in 17 *ACTN3* and 19 *ACE* studies being judged as adequately meeting the 159 predetermined inclusion criteria and subsequently being included in the final analysis (see Figure 1 for 160 full systematic search process).

161

****Insert Figure 1. near here****

162 **3.2 Study Characteristics**

163 The 17 ACTN3 studies consisted of nine case-control and eight cohort studies respectively. There were 164 a total of 1759 football players included across all studies (aged 10-37 years), with sample sizes ranging 165 from 25-353. The most frequently studied nationality was Brazilian (n = 850), whilst the most 166 frequently studied ethnicity was Caucasian (n = 587). Eleven studies included PRO players (n = 713; 167 aged 17-37 years); four studies included NP players (n = 447; aged 14-30 years); and, two included both PRO and NP players (n = 599; aged 10-27 years). The 19 ACE studies consisted of 15 case-control 168 169 and four cohort studies. There were a total of 1925 football players included across all studies (aged 10-170 27 years), with sample sizes ranging from 25-353. The most frequently studied nationality was Brazilian 171 (n = 709), whilst the most frequently studied ethnicity was Caucasian (n = 802). Ten studies included PRO players (n = 674; aged 17-26 years); seven studies included NP players (n = 652; aged 15-21 172 173 years); and, two included both PRO and NP players (n = 599; aged 10-27 years).

174 **3.3 Risk of Bias**

175 Fifteen of the originally included 17 ACTN3 studies remained after risk of bias assessment. The studies excluded, with reasons for exclusion, were as follows: (1) Massidda et al. ⁵⁷ failed HWE assessment; 176 and, (2) La Montagna et al. ⁵⁸ failed to provide the specific nationality and ethnicity of the footballers, 177 178 thus an ethnically-matched CON could not be added. Seventeen of the originally included 19 ACE 179 studies remained after risk of bias assessment. The following studies were excluded due to failing HWE assessment: Gineviciene et al. 59 and Galeandro et al. 41. In addition, the Bulgarian sub-cohort of 180 Andreeva et al. 60 was excluded from analysis due to also failing HWE assessment. In all genetic 181 182 comparison models that identified a significant association with ACTN3 and ACE, funnel plots did not 183 reveal any signs of asymmetry and Egger's test did not detect a significant indication of publication 184 bias.

185 **3.4 Main Analysis**

186 Several significant associations were observed between the ACTN3 R577X polymorphism and genetic 187 comparison models with case-control studies in isolation: (1) allele contrast; (2) recessive; (3) 188 dominant; (4) RR vs. XX; and, (5) RR vs. RX. In addition, similar associations with the same genetic 189 models were also observed in case-control studies combined with cohort studies; with the only 190 exception being the additional significant association of RX vs. XX. The strongest association observed 191 in both case-controls in isolation and combined with cohorts was RR vs. XX (see Table 1 for ORs and 192 CIs of each ACTN3 comparison). Sensitivity analysis assessed the robustness of the results and revealed 193 that no study significantly altered pooled ORs. Conversely, no significant associations were observed 194 between the ACE I/D polymorphism with case-controls in isolation or combined with cohorts using any 195 genetic comparison model (see Table 2 for ORs and CIs of each ACE comparison).

196 3.4.1 Sub Analyses

197 The first sub-analysis assessed the independent associations of PRO players and NP players vs. controls 198 (CON). Several significant associations were observed between the ACTN3 R577X polymorphism and 199 genetic comparison models in PROs: (1) allele contrast; (2) recessive; (3) dominant; (4) RR vs. XX; 200 and, (5) RR vs. RX. The strongest association observed was RR vs. XX. No significant associations 201 were observed between the ACTN3 R577X polymorphism and genetic comparison models in NPs vs. 202 CON. Conversely, no significant associations were observed between the ACE I/D polymorphism and 203 PRO players. However, several significant associations were observed between the ACE I/D 204 polymorphism and NP players: (1) allele contrast; (2) dominant; and, (3) ID vs. DD.

The second sub-analysis assessed the independent influence of ethnicity and nationality on associations. Due to the variance in geographical ancestry, Caucasian and Brazilian were the only ethnicity and nationality eligible for analysis. Several significant associations were observed between the *ACTN3* R577X polymorphism and genetic comparison models in Caucasians: (1) allele contrast;

- (2) recessive; (3) dominant; and, (4) RR vs. XX. Likewise, several significant associations were also
 observed between the *ACTN3* R577X polymorphism and genetic comparison models in Brazilians: (1)
- allele contrast; (2) dominant; (3) RR vs. XX; and, (4) RX vs. XX. The strongest association observed
- 212 in both Caucasians and Brazilians was RR vs. XX. Conversely, no significant associations were
- 213 observed between the ACE I/D polymorphism and Caucasians or Brazilians.
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****Insert **Table 1.** near here****

****Insert **Table 2.** near here****

216 **4 Discussion**

The aim of this study was to assess if the *ACTN3* R577X and/or *ACE* I/D polymorphism were associated with athlete status in football; and if so, determine which allele and/or genotypes are most likely to influence this phenotype via a meta-analysis. To the author's knowledge this is the first review to do this within football; and moreover, it is the first meta-analysis in a homogenous team-sport cohort.

221 4.1 Main Analysis

222 Following meta-analysis, this review identified several associations between the ACTN3 223 R577X polymorphism and all football players (PRO & NP combined) vs. CON. In summary, the main 224 associations were observed in football players possessing the R allele, with the strongest association 225 being the RR vs. XX genotype. These associations were observed in case-control studies in isolation 226 and remained similar with the addition of cohort studies. Although, with the significant increase in 227 sample size from cohort studies the CI was reduced, possibly indicating a more accurate estimation. 228 The only notable difference between genetic comparison models was that the RX vs. XX pair-wise 229 comparison was statistically significant with cohorts added vs. non-significant in case-controls in 230 isolation. Therefore, only the over-dominant model remained non-significant in both case-control 231 studies in isolation and with the addition of cohort studies; most likely due to the strength of the 232 association of the RR genotype. As such, the results of this analysis showcase that in football players, 233 similar to power athletes, there is an overrepresentation of the ACTN3 RR genotype. For instance, this 234 can be illustrated by the similar recessive model and dominant model findings of the present study and 235 Ma and colleagues ¹⁶, respectively. This can likely be explained by the combination of several factors: 236 (1) the number and frequency of powerful actions performed in a game (i.e., 1000-1400 acyclical bursts 237 of activity, including; jumps, tackles, shots at goal, changes of direction, and, sprints), with a high-238 intensity sprint occurring every \sim 70s ^{61,62}; (2) the contribution of the ability to repeat higher intensity actions to success in football (i.e., league position, goals scored, goals prevented, duel success) ^{63,64}; 239 240 and, (3) the association of the RR genotype with power-orientated phenotypes (i.e., vertical jumping and 10-30m sprints) ^{65,66}. However, it is important to recognise that there are many other genetic 241 polymorphisms, with each likely influencing performance to a limited extent ^{45,46}. Therefore, football 242

243 players, power athletes, and indeed other team-sports, may possess contrasting allele and genotype 244 frequencies in other polymorphisms. Hence, the results of this review only indicate a difference between 245 football players and non-athletic CONs. However, this difference appears to be greatly mediated by 246 competitive playing level, which may have contributed to the heterogeneity present between ACE 247 studies; and as a result, the non-significant associations between ACE and PRO & NP players combined. 248 Indeed, the results of the subsequent sub-analysis on competitive playing level revealed that ACE may 249 play an important role in determining athlete status earlier in athlete development. Therefore, although 250 ACE was not associated with athlete status in the main analysis, it is still an important genetic variant 251 in football, depending on specific competitive playing levels.

252 4.2 Sub-Analyses

253 The importance of the competitive playing level of footballers in this review was assessed via 254 separating PRO players and NP players. PRO players were classified as players that studies specifically 255 described as playing at a professional level, whereas NP players were classified as players playing at a 256 semi-professional, amateur, or youth level. This particular method of categorisation was chosen, as 257 opposed to elite vs. non-elite, to circumvent the problematic classification issue of 'eliteness' (i.e., what constitutes elite status and how do we define it?)^{67,68}. For example, players may be internationals who 258 259 represent their country (normally classed as elite), however what if their country is near the bottom of 260 the international rankings? Furthermore, players may play for a European club positioned 1st in their 261 country's highest league (normally classed as elite), but what if the country and club both have a low 262 UEFA coefficient ranking? Moreover, how do we define the highest performing youth players? Given 263 that no irrefutable solution has been provided and no general consensus has been agreed, the term 'elite' 264 is still inconsistently utilised in the literature to describe varying standards of performance ^{67,68}. 265 Therefore, to reduce between-study-heterogeneity, the authors chose to compare PROs and NPs 266 (limitations of this approach are discussed in section 4.3).

267 Firstly, the results of this analysis revealed that the ACTN3 R577X polymorphism was associated with PRO players vs. CON. However, no statistically significant difference was observed 268 269 between NP players and CON. Moreover, the strength of the association between the R allele, and 270 specifically the RR genotype, was even stronger in PRO players when separated from NP players. As 271 such, these results indicate that the R allele is a likely (albeit small) contributing factor towards attaining 272 PRO status in football. Secondly, whilst the ACE I/D polymorphism was not associated with PRO 273 players vs. CON, it was associated with NP players vs. CON. To be specific, NP players were more 274 likely to possess a D allele or DD genotype. This is perhaps more easily demonstrated via inverse 275 statistical analysis: (1) D vs. I (OR = 1.18, 95% CI: 1.01-1.38); (2) DD vs. DI+II (OR = 1.29, CI: 1.02-276 1.63); and, (3) DD vs. DI (OR = 1.32, CI: 1.03-1.69). Interestingly, the NP players consisted solely of 277 youth players in the ACE I/D studies (n = 652; aged 15-21 years) showcasing that the D allele is only 278 overrepresented specifically in youth football. Furthermore, the differences between PROs and NPs

- may have increased further given that Egorova and colleagues ⁴² was not included in this analysis, due 279 280 to not displaying the individual genotype frequencies of PRO and NP players in their study. However, 281 the authors did report that when separated by competitive playing level, only elite players displayed a 282 significantly higher frequency of the ACTN3 R allele compared to CON (81.1%, P < 0.001); whilst 283 conversely, only youth players displayed a significantly higher frequency of the ACE D allele compared 284 to CON (78.8%, P < 0.001). As such, the observed cumulative evidence suggests that whilst possessing the ACE D allele is potentially more beneficial to young players, the ACTN3 R allele is the only allele 285 286 which likely has a (minor) role in attaining PRO status (see Figure 2). However, it is important to note 287 the use of terms such as "albeit small" and "minor" when interpreting these findings. Polymorphisms 288 account for very little of the inter-individual variance in complex traits such as athlete status ⁴⁵. Indeed, even a combination of 97 polymorphisms (at $P = 5 \times 10^{-8}$ or better) could only account for 2.7% of body 289 mass index variance between individuals⁶⁹. Therefore, as evidenced by the relatively small odds ratios 290 291 in this study, ACTN3 and ACE similarly play a minor role in determining athlete status in football.
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****Insert Figure 2. near here****

293 The explanation for the observed associations between the ACE D allele and ACTN3 R allele 294 with youth and PRO players respectively is challenging. Both alleles have been previously associated 295 with strength/power orientated sports and general strength/power characteristics (i.e., increased percentage of type II muscle fibres; strength; and, muscle mass) ^{13,23}. More specifically, in football 296 297 cohorts both alleles have been positively associated with greater countermovement and squat jump 298 performance, and faster 10m, 20m, and 30m sprint times ^{65,66}. As such, both alleles appear to be 299 associated with the same side of the endurance-power continuum in football. Therefore, it is interesting 300 that each allele is independently associated with different competitive playing levels. However, perhaps 301 the categorisation method employed to distinguish competitive playing levels in this review could 302 possibly be responsible. For example, whilst heterogeneity between studies in the PRO vs. NP analysis 303 was mostly small, the age of NPs in ACTN3 ranged from 14-30 years; whereas the age of NP players 304 in ACE ranged from 15-21 years. As such, it is possible that the NP ACE players had greater 305 performance levels than the NP ACTN3 players, as the NP ACE players may play at the highest youth 306 level in their respective age groups; whilst the ACTN3 players may not. Indeed, several previous studies have demonstrated that 'elite' youth players, aged 14-17 years, have outperformed non-elite players in 307 acceleration, speed and jumping assessments ^{70,71}. Although, it could be argued that grouping young 308 309 players (14-17 years) as elite and non-elite is not appropriate as factors such as maturation status may 310 influence performance more than ACTN3 or ACE genotype. However, in truth, numerous potential 311 explanations exist, including: (1) differences in competitive performance levels; (2) variations in 312 geographical ancestry; (3) disparities in the distributions of players relative to their on-field position 313 within samples; (4) distinct methods of genotyping; (5) individual gene-gene interactions; and, (6) 314 separate gene-environment interactions. Moreover, it is also important to note that the ACE gene is part of a very complex pathway, with serval interactions that can influence it's activity; whilst *ACTN3* represents the presence/absence of a structural protein ^{18,19}. As such, the specific cause of the distinction between the *ACTN3* R577X and *ACE* I/D polymorphisms with PRO and NP players requires further research to elucidate these findings.

319 In addition to competitive playing level, sub-analysis also assessed the influence of ethnicity 320 and geographical ancestry on observed associations. Therefore, studies, and samples within studies, 321 were separated based upon their reported ethnic heritage and nationality. After separation, Caucasians 322 and Brazilians collectively represented 81% and 79% of ACTN3 and ACE studies respectively. 323 Therefore, only these could be assessed via comparison analysis. Firstly, in relation to the ACE I/D 324 polymorphism, no significant associations were observed in either Caucasians or Brazilians. However, 325 in relation to the ACTN3 R577X polymorphism, several significant associations were observed in both 326 Caucasians and Brazilians between the R allele and football players vs. CON. However, in models with 327 significant associations in both Caucasians and Brazilians (allele contrast, dominant, RR vs. XX), Caucasians displayed higher ORs. This suggests the R allele, and more significantly the RR genotype, 328 329 appear to be of greater importance to footballers of Caucasian heritage. This suggestion is bolstered by 330 the significant associations observed solely between Caucasians and a recessive model, and between 331 Brazilians and the pair-wise comparison of the RX vs. XX genotypes. This reveals that the RX genotype 332 may be more associated with footballing status in Brazilians, whilst the RR genotype may be more 333 associated with footballing status in Caucasians.

334 The possible cause of the disparities in ORs between the R allele in Brazilians and Caucasians 335 may be related to the ethnic diversity of the Brazilian population; and more specifically, the proportion 336 of individuals with African ancestry. Individuals of African ancestry have greater frequencies of the 337 RR genotype (~78%) and significantly lesser frequencies of the XX genotype (~1%), irrespective of athlete status, compared to the estimated frequencies of the world population (RR ~40%; XX ~18%)⁷²⁻ 338 ⁷⁴. As such, the ACTN3 R577X polymorphism does not differentiate elite athletes from CON of African 339 ethnicity in either power or endurance-orientated sports ^{72,73}. Using ancestry informative markers, 340 previous studies have reported that the Brazilian population are formed of ~65% European SNPs, ~22% 341 342 African SNPs, and ~13% Amerindian SNPs⁷⁵. Furthermore, these proportions vary in each independent region of Brazil. For example, it is estimated that the population of Rio de Janeiro is formed of ~55% 343 European SNPs, ~31% African SNPs, and ~14% Amerindian SNPs⁷⁶. As such, given the genetic 344 similarity of the Brazilian population with the genetic profile of African ethnicities, this potentially 345 346 explains why the RR genotype may contribute to footballing status in Brazil to a lesser extent.

347 4.3 Strengths and Limitations

348 This study's findings are reinforced due to the small-moderate heterogeneity identified between 349 studies. Only two of the 28 total measurements regarding *ACTN3* displayed an I^2 value over 50% and violated Cochran's Q statistical threshold (P<0.05); although neither of these two measurements displayed a statistically significant result. Likewise, neither of the three associations identified between the *ACE* I/D polymorphism and NP players displayed significant heterogeneity. Furthermore, no suggestion of publication bias was identified regarding the observed significant associations, through analysis of funnel plots or Egger's test (P<0.05).

355 This review is not without limitations and each must be duly considered when interpreting the 356 findings. Firstly, this review attempted to include both male and female football players, but unfortunately the authors could only locate one study involving female players ⁶⁰ which met the pre-357 358 defined inclusion criteria. Therefore, the results from this review mainly apply to male football players. 359 Secondly, the ethnic and geographic implications of this review mainly concern Brazilians and 360 Caucasians, due to both representing the majority of the football players included in this review (ACTN3 361 = 48% & 33% respectively; ACE = 37% & 42% respectively). Thus, research is still required on players 362 of diverse geographical ancestry. Thirdly, the division of players into PRO and NP categories was 363 chosen to circumvent the well-known issues surrounding what constitutes 'eliteness' in sport; however, 364 the PRO level still encompasses substantial disparities in standards of play (e.g., English Premier 365 League vs. English Football League Two). As such, this makes it difficult to objectively quantify player 366 ability and consequently associate the findings with specific levels of performance. Finally, the on-field 367 positions of footballers were rarely reported in the included studies. This is important considering the 368 previously reported inter-positional associations between the ACTN3 R577X and ACE I/D 369 polymorphism and football players, such as forwards and goalkeepers possessing significantly higher frequencies of the ACTN3 R allele and ACE D allele, respectively 42. Therefore, it is unknown whether 370 371 the associations in this review were influenced by the number of players included in each study relative 372 to their on-field positions.

373 4.4 Practical Applications

374 Limited evidence exists supporting the practical application of genetic information with athlete 375 development. For example, it has been reported that individuals possessing the ACTN3 RR and ACE 376 DD genotypes may have an improved adaptive response in strength and power with heavy resistance training ^{77,78}. As such, Kikuchi and Nakazato ⁷⁹ suggested that individuals possessing power-orientated 377 genetic variants, such as ACTN3 RR, may benefit more from resistance training consisting of high 378 weights with low-repetitions for strength and power. Jones and colleagues 80 investigated this theory 379 380 and reported that in 39 football players the ACE DD and ACTN3 RR genotypes demonstrated a greater 381 improvement in countermovement jump performance in response to high-weight low-repetition 382 resistance training; whereas, the ACE II and ACTN3 XX genotypes demonstrated a greater improvement 383 with low-load high-repetition resistance training. However, this study comprised a small sample size, 384 which consequently increases the likelihood of type 1 error occurrence⁸¹. Furthermore, to the author's knowledge, the results presented by Jones and colleagues ⁸⁰ have yet to be replicated in an independent 385

- 386 cohort. Therefore, further intervention studies are required in order to establish a strong evidence base, 387 before practical recommendations can be proposed regarding the results of this review. However, once 388 a strong evidence base has been established which supports genetic-based programme design, this 389 review can be used alongside a number of other extensively evidenced/researched genotypes to form 390 the basis of genetic-based training. Indeed, studies have also showcased that differences in biomarkers 391 of muscle damage, hormones, and inflammatory responses, may be influenced by genotype variation 392 ^{82,83}. Therefore, the utilisation of genetic information in the future may not only aid in optimising
- 393 training adaptation, but also the planning of athlete workloads and recovery.

394 5 Conclusion

395 The results of this review provide further evidence that individual genetic variation likely contributes 396 towards athlete status. Our findings suggest that genetic variation can differentiate athletes of different 397 competitive playing statuses in a homogenous team-sport cohort. Specifically, this review has 398 showcased that the R allele and RR genotype of the ACTN3 R577X polymorphism are overrepresented 399 in PRO football players; whereas the D allele and DD genotype of the ACE I/D polymorphism are 400 overrepresented in youth players. These overrepresentations may be explained by the association of the 401 ACTN3 RR genotype and ACE DD genotype with power-orientated phenotypes and the relative 402 contribution of these power-orientated phenotypes to success in football. As such, the ACTN3 R577X and ACE I/D polymorphisms are likely (albeit relatively minor) contributing factors which influence 403 404 athlete status in football.

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Figure 1. Flow diagram of systematic search process.

			Case-Cor	ntrol					
Model			n test	Heterogeneity			eity	Bias	
NIOG	OR	95% CI	Р		P		I^2	Egger's	
Allele co	ontrast	1.30	1.15-1.46	< 0.001			0.68	0.00	0.88
Recess	sive	1.42	1.20-1.68	< 0.001			0.22	0.24	0.87
Domir	ant	1.35	1.06-1.72	0.016			0.33	0.13	0.97
Over-dor	ninant	0.84	0.66-1.07		0.16		0.05	0.45	0.93
RR vs.	XX	1.67	1.28-2.17	< 0.001			0.63	0.00	0.98
RR vs.	RX	1.38	1.07-1.78	0.013			0.08	0.41	0.76
RX vs.	XX	1.17	0.90-1.51		0.23		0.11	0.38	0.92
		(Case-Control	& Cohort					
Allele co	ontrast	1.26	1.15-1.38	< 0.001			0.72	0.00	0.48
Recess	sive	1.31	1.15-1.50	< 0.001			0.23	0.18	0.89
Domir	ant	1.40	1.17-1.69	< 0.001			0.49	0.00	0.64
Over-dor	ninant	0.93	0.78-1.12		0.45		0.05	0.38	0.75
RR vs.	XX	1.60	1.31-1.96	< 0.001			0.72	0.00	0.62
RR vs.	RX	1.23	1.02-1.49	0.029			0.08	0.34	0.83
RX vs.	XX	1.29	1.06-1.57	0.010			0.18	0.24	0.69
			PRO vs.	NP					
	PRO	1.35	1.18-1.53	< 0.001			0.77	0.00	0.34
Allele contrast	NP	1.07	0.90-1.27		0.44		0.90	0.00	0.30
D .	PRO	1.48	1.23-1.77	< 0.001			0.30	0.14	0.70
Recessive	NP	1.00	0.78-1.29		0.97		0.74	0.00	0.46
	PRO	1.40	1.09-1.81	0.010			0.16	0.30	0.83
Dominant	NP	1.27	0.91-1.76		0.16		1.00	0.00	0.34
	PRO	0.84	0.63-1.12		0.24	0.020		0.51	0.89
Over-dominant	NP	1.13	0.89-1.43		0.33		0.79	0.00	0.66
DD VV	PRO	1.77	1.34-2.34	< 0.001			0.57	0.00	0.61
KK VS. XX	NP	1.22	0.84-1.76		0.29		0.95	0.00	0.21
	PRO	1.41	1.07-1.86	0.014			0.07	0.40	0.98
KK VS. KA	NP	0.94	0.72-1.23		0.68		0.72	0.00	0.52
	PRO	1.22	0.79-1.87		0.36	0.028		0.50	0.91
KX vs. XX	NP	1.30	0.92-1.84		0.14		0.99	0.00	1.00
			Caucasian & I	Brazilian				•	
A 11-1	Caucasian	1.32	1.14-1.53	< 0.001			0.39	0.05	0.73
Allele contrast	Brazilian	1.23	1.07-1.41	0.003			0.45	0.00	0.47
D	Caucasian	1.41	1.14-1.74	0.001			0.11	0.42	0.67
Recessive	Brazilian	1.19	0.98-1.44		0.07		0.64	0.00	0.69
	Caucasian	1.49	1.11-2.00	0.008			0.79	0.00	0.82
Dominant	Brazilian	1.52	1.15-1.99	0.003			0.17	0.35	0.92
	Caucasian	0.91	0.75-1.12		0.38		0.11	0.42	0.69
Over-dominant	Brazilian	1.07	0.89-1.29		0.48		0.59	0.00	0.45
	Caucasian	1.71	1.24-2.36	0.001			0.60	0.00	0.60
KK VS. XX	Brazilian	1.61	1.19-2.16	0.002			0.23	0.27	0.94
	Caucasian	1.36	0.96-1.92		0.08		0.08	0.47	0.61
KK vs. KX	Brazilian	1.08	0.88-1.32		0.46		0.66	0.00	0.85
	Caucasian	1.34	0.98-1.83		0.06		0.61	0.00	0.97
ΚΛ VS. ΛΛ	Brazilian	1.48	1.10-1.98	0.009			0.17	0.35	0.98

Table 1. Statistical analysis of all studies investigating the ACTN3 R577X polymorphism (n = 15)

Note. Allele contrast (R vs. X); Recessive (RR vs. RX+XX); Dominant (RR+RX vs. XX); Over-dominant (RX

vs. RR+XX); OR = Odds Ratio; CI = Confidence Interval; PRO = Professional; NP = Non-Professional.

			Case-Con	trol		r			
Model		Association test				Hete	Bias		
IVIOU	el	OR	95% CI	Р		Р		I^2	Egger's
Allele co	ontrast	0.88	0.73-1.07		0.21	0.001		0.61	0.16
Recess	sive	0.84	0.69-1.03		0.09		0.44	0.01	0.72
Domir	nant	0.84	0.62-1.15		0.28	< 0.001		0.66	0.09
Over-dor	ninant	0.93	0.74-1.16		0.51	0.040		0.44	0.04
II vs.]	DD	0.82	0.58-1.17		0.27	0.025		0.47	0.19
II vs.	ID	0.98	0.79-1.21		0.82		0.65	0.00	0.27
ID vs.	DD	0.84	0.62-1.14		0.26	0.002		0.60	0.08
		(Case-Control &	& Cohort					
Allele co	ontrast	0.87	0.75-1.02		0.09	0.004		0.53	0.10
Recess	sive	0.86	0.72-1.03		0.10		0.69	0.00	0.69
Domir	nant	0.82	0.63-1.07		0.14	< 0.001		0.62	0.05
Over-dor	ninant	0.90	0.73-1.10		0.31	0.019		0.45	0.03
II vs.]	DD	0.79	0.60-1.05		0.10		0.08	0.34	0.12
II vs.	ID	1.00	0.83-1.22		0.97		0.75	0.00	0.21
ID vs.	DD	0.81	0.62-1.06		0.13	0.001		0.58	0.05
			PRO vs.	NP					
	PRO	1.01	0.86-1.19		0.88		0.96	0.00	0.67
Allele contrast	NP	0.85	0.73-0.99	0.044			0.15	0.35	0.45
D .	PRO	0.97	0.72-1.29		0.82		0.99	0.00	0.99
Recessive	NP	0.88	0.65-1.18		0.39		0.20	0.29	0.38
D	PRO	1.05	0.82-1.34		0.71		0.86	0.00	0.73
Dominant	NP	0.78	0.61-0.98	0.032			0.25	0.22	0.44
	PRO	1.07	0.85-1.34		0.57		0.81	0.00	0.62
Over-dominant	NP	0.86	0.69-1.07		0.18		0.22	0.27	0.95
и рр	PRO	1.05	0.74-1.49		0.77		0.99	0.00	0.95
II vs. DD	NP	0.77	0.55-1.08		0.14		0.22	0.26	0.26
и п	PRO	0.94	0.69-1.27		0.69		0.98	0.00	0.85
II vs. ID	NP	0.97	0.71-1.33		0.85		0.20	0.28	0.51
	PRO	1.05	0.80-1.36		0.73		0.80	0.00	0.55
ID vs. DD	NP	0.76	0.59-0.97	0.030			0.38	0.07	0.56
		(Caucasian & E	Brazilian					
A 11 1	Caucasian	0.81	0.57-1.14		0.23	0.004		0.74	0.04
Allele contrast	Brazilian	0.85	0.70-1.03		0.09		0.37	0.07	0.63
D .	Caucasian	0.75	0.57-1.00		0.05		0.61	0.00	0.50
Recessive	Brazilian	0.83	0.60-1.14		0.25		0.25	0.26	0.07
	Caucasian	0.75	0.42-1.34		0.34	< 0.001		0.82	0.03
Dominant	Brazilian	0.77	0.55-1.06		0.11		0.17	0.38	0.73
	Caucasian	0.83	0.53-1.29		0.40	0.009		0.70	0.08
Over-dominant	Brazilian	1.05	0.64-1.71		0.85	0.018		0.67	0.29
	Caucasian	0.65	0.38-1.10		0.11		0.08	0.52	0.08
II vs. DD	Brazilian	0.80	0.52-1.22		0.29		0.31	0.17	0.58
	Caucasian	1.05	0.77-1.44		0.75		0.59	0.00	0.28
II vs. ID	Brazilian	0.88	0.63-1.23		0.45		0.10	0.48	0.19
10 55	Caucasian	0.76	0.41-1.42		0.39	< 0.001		0.82	0.07
ID vs. DD	Brazilian	0.82	0.47-1.42		0.48		0.09	0.51	0.57

Table 2. Statistical analysis of all studies investigating the *ACE* I/D polymorphism (n = 17)

Note. Allele contrast (I vs. D); Recessive (II vs. ID+DD); Dominant (II+ID vs. DD); Over-dominant (ID vs.

II+DD); OR = Odds Ratio; CI = Confidence Interval; PRO = Professional; NP = Non-Professional.

	Experim	ental	Co	ntrol		Odds Ratio			
Study	Events	Total	Events	Total			OR	95%-Cl	W(fixed)
		100	100				1.00	14 00 0 FF1	0.40/
Santiago et al. (2008)	80	120	136	246			1.62	[1.03; 2.55]	8.1%
Pimenta et al. (2012)	260	400	920	1820			1.09	[0.00, 1.74]	32.2%
Ulucan et al. (2015)	31	50	38	60			0.94	[0.43: 2.05]	2.8%
Galeandro et al. (2017)	59	86	152	256			1.50	[0.89; 2.51]	6.2%
Honarpour et al. (2017)	107	180	215	400		- <u>-</u>	1.26	[0.88; 1.80]	13.1%
Salgueirosa et al. (2017)	56	86	249	412			1.22	[0.75; 1.99]	7.1%
Coelho et al. (2018)	116	166	126	200			1.36	[0.88; 2.11]	8.7%
Clos et al. (2019)	31	46	110	226			2.18	[1.12; 4.26]	3.7%
Clos et al. (2019)_1	17	26	120	214			1.48	[0.63; 3.47]	2.3%
Clos et al. (2019)_2	12	14	207	226			0.55	[0.11; 2.64]	0.7%
Cocci et al. (2019)	62	110	112	204			1.06	[0.67; 1.69]	1.1%
Fixed effect model		1358		6008		4	1.35	[1.18: 1.53]	100%
Heterogeneity: I-squared=09	%, tau-squa	red=0,	p=0.7739					[,]	
(A)									
(A)					0.2	0.5 1 2 5	5		
			•			Odde Patio			
Ctudy	Experim	ental	Co	ntrol				05% 01	W/(fixed)
Sludy	Events	Total	Events	rotai			OR	95%-CI	w(lixea)
Santiago et al. (2008)	29	60	35	123		-	2 35	[1 24 4 46]	81%
Pimenta et al. (2012)	15	37	319	964			1.38	[0 71 2 69]	7 4%
Pimenta et al. (2013)	82	200	272	818			1.39	[1 02: 1 92]	32.9%
Ulucan et al. (2015)	11	25	15	30			0.79	[0.27: 2.28]	2.9%
Galeandro et al. (2017)	21	43	40	128			2.10	[1.04: 4.25]	6.6%
Honarpour et al. (2017)	37	90	50	200		<u> </u>	2.09	[1.24: 3.55]	11.9%
Salqueirosa et al. (2017)	18	43	76	206			1.23	[0.63: 2.40]	7.4%
Coelho et al. (2018)	36	83	40	100			1.15	[0.64; 2.07]	9.5%
Clos et al. (2019)	9	23	22	113		*	- 2.66	[1.02; 6.93]	3.6%
Clos et al. (2019)_1	6	13	33	107			1.92	[0.60; 6.16]	2.4%
Clos et al. (2019)_2	5	7	94	113			0.51	[0.09; 2.80]	1.1%
Cocci et al. (2019)	15	55	32	102			0.82	[0.40; 1.70]	6.3%
Fired offerstrees del		670		0004			4 40	[4 00. 4 77]	1000/
Fixed effect model	13% tau-s	679	-0.0187 n	3004 -0 3046		- \$	1.48	[1.23; 1.77]	100%
Fixed effect model Heterogeneity: I-squared=14	1.3%, tau-so	679 quared	=0.0187, p	3004 =0.3046	;		1.48	[1.23; 1.77]	100%
Fixed effect model Heterogeneity: I-squared=14 (B)	l.3%, tau-so	679 guared	=0.0187, p	3004 = <i>0.3046</i> C).1	0.5 1 2	1.48	[1.23; 1.77]	100%
Fixed effect model Heterogeneity: I-squared=14 (B)	1.3%, tau-so	679 quared	=0.0187, p	3004 <i>=0.3046</i> C	5 	0.5 1 2	1.48	[1.23; 1.77]	100%
Fixed effect model Heterogeneity: I-squared=14 (B)	1.3%, tau-so Experim	679 guared	=0.0187, p	3004 <i>=0.3046</i> ℃).1	0.5 1 2 Odds Ratio	1.48	[1.23; 1.77]	100%
Fixed effect model Heterogeneity: I-squared=14 (B) Study	1.3%, tau-so Experim Events	679 guared dental Total	=0.0187, p. Co Events	3004 <i>=0.3046</i> C Ontrol Total	5 	0.5 1 2 Odds Ratio	1.48 10 OR	[1.23; 1.77] 95%-Cl	100% W(fixed)
Fixed effect model Heterogeneity: I-squared=14 (B) Study	1.3%, tau-so Experim Events	679 guared iental Total	=0.0187, p. Co Events	3004 =0.3046 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	5 	0.5 1 2 Odds Ratio	1.48	[1.23; 1.77] 95%-Cl	100% W(fixed)
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003)	Experim Events	679 guared nental Total 150	=0.0187, p. Co Events	3004 =0.3046 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	5 	0.5 1 2 Odds Ratio	1.48 10 OR 0.68	[1.23; 1.77] 95%-Cl [0.41; 1.15]	100% W(fixed) 9.2%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Androaue at al. (2011)	8.3%, tau-so Experim Events 86 180	679 guared nental Total 150 250	=0.0187, p. Co Events 69 200	3004 =0.3046 0 ontrol Total 104 304 268	5 	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76: 0.57]	100% W(fixed) 9.2% 18.7% 15%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Avad et al. (2014)	8.3%, tau-so Experim Events 86 180 15 86	679 guared nental Total 150 250 18 136	=0.0187, p Co Events 69 200 174 127	3004 =0.3046 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	5 	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55]	100% W(fixed) 9.2% 18.7% 1.5% 12.1%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014)	Experim Events 86 180 15 86 281	679 guared tental Total 150 250 18 136 520	=0.0187, p Co Events 69 200 174 127 109	3004 =0.3046 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	5 []).1	0.5 1 2 Odds Ratio	1.48 10 0.68 1.34 2.70 0.99	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Jones et al. (2016)	5.3%, tau-so Experim Events 86 180 15 86 281 44	679 guared tental Total 150 250 18 136 520 78	=0.0187, p Co Events 69 200 174 127 109 88	3004 =0.3046 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	5 []).1	0.5 1 2 Odds Ratio	1.48 10 0.68 1.34 2.70 0.99 0.98 1.38	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Jones et al. (2016) Jones et al. (2017)	2.3%, tau-so Experim Events 86 180 15 86 281 44 281	679 guared tental Total 150 250 18 136 520 78 438	=0.0187, p Cc Events 69 200 174 127 109 88 80	3004 =0.3046 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	5 D.1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36] [0.98; 2.10]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Jones et al. (2016) Jones et al. (2017) Menezes et al. (2019)	2.3%, tau-so Experim Events 86 180 15 86 281 44 281 67	679 guared 150 250 18 136 520 78 438 130	=0.0187, p Cc Events 69 200 174 127 109 88 80 48	3004 =0.3046 Ontrol Total 104 304 268 200 200 182 144 120	5 D.1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36] [0.98; 2.10] [0.97; 2.63]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019)	2.3%, tau-so Experim Events 86 180 15 86 281 44 281 67	679 guared 150 250 18 136 520 78 438 130	=0.0187, p. Cc Events 69 200 174 127 109 88 80 48	3004 =0.3046 Ontrol Total 104 304 268 200 200 182 144 120	5 []).1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36] [0.98; 2.10] [0.97; 2.63]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model	2.3%, tau-so Experim Events 86 180 15 86 281 44 281 67	679 guared Total 150 250 18 136 520 78 438 130 1720	=0.0187, p. Events 69 200 174 127 109 88 80 48	3004 =0.3046 Control Total 104 304 268 200 200 182 144 120 1522	5 [] D.1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.8; 2.10] [0.97; 2.63] [1.01; 1.38]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3	2.3%, tau-so Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-so	679 guared Total 150 250 18 136 520 78 438 130 1720 guared	=0.0187, p. Events 69 200 174 127 109 88 80 48	3004 =0.3046 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	2	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.8; 2.10] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=33 (C)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-so	679 guared Total 150 250 18 136 520 78 438 130 1720 guared	=0.0187, p. Events 69 200 174 127 109 88 80 48	3004 c ontrol Total 104 304 200 200 182 200 182 144 120 1522 <i>=0.1432</i>	2 .1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se	679 guared Total 150 250 18 136 520 78 438 130 1720 guared	=0.0187, p. Cc Events 69 200 174 127 109 88 80 48	3004 =0.3046 C Dontrol Total 104 304 200 200 182 144 120 1522 =0.1432	2 0.1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.8; 2.10] [0.97; 2.63] [1.01; 1.38]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-so Experim	679 guared Total 150 250 18 136 520 78 438 130 1720 guared ental	=0.0187, p. Cc Events 69 200 174 127 109 88 80 48	3004 =0.3046 C Dontrol Total 104 304 200 200 182 144 120 1522 =0.1432	2 0.1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-so Experim Events	679 guared Total 150 250 18 136 520 78 438 130 1720 guared ental Total	=0.0187, p. Cc Events 69 200 174 127 109 88 80 48 2=0.0298, p Co Events	3004 =0.3046 C Dontrol Total 104 304 200 200 182 144 120 1522 =0.1432 ntrol Total	2 0.1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 OR	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed)
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-so Experim Events	679 guared Total 150 250 18 136 520 78 438 130 1720 guared ental Total	=0.0187, p. Cc Events 69 200 174 127 109 88 80 48 2=0.0298, p Co Events	3004 =0.3046 C Dontrol Total 104 304 200 200 182 144 120 1522 =0.1432 ntrol Total	2 0.1	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 OR	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 23.1% 16.9% 9.8% 100% W(fixed)
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events	679 guared Total 150 250 18 136 520 78 438 130 1720 guared ental Total	=0.0187, p. Co Events 69 200 174 127 109 88 80 48 -0.0298, p Co Events	3004 =0.3046 C pontrol Total 104 304 200 200 182 144 120 1522 =0.1432 ntrol Total	2 0.1	0.5 1 2 Odds Ratio 	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 OR 0.74	95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011)	2.3%, tau-so Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-so Experim Events 32 64	679 guared Total 1500 2500 18 136 5200 78 438 1300 1720 guared ental Total	=0.0187, p. Cc Events 69 200 174 127 109 88 80 48 48 48 48 48 48 48 48 48 48 48 48 48	3004 =0.3046 C ontrol Total 104 268 200 200 182 144 120 1522 =0.1432 ntrol Total	2 0.1 0.2	0.5 1 2 Odds Ratio 	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 OR 0.74 1.33	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.33; 2.14]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 6 2	679 guared Total 150 250 18 136 520 78 438 130 1720 quared ental Total 75 125 9 9	=0.0187, p. Cc Events 69 200 174 127 109 88 80 48 200 174 127 109 88 80 48 200 200 200 200 200 200 200 200 200 20	3004 =0.3046 C ontrol Total 104 268 200 200 182 144 120 1522 =0.1432 ntrol Total 52 152 134	2 0.1 0.2	0.5 1 2 Odds Ratio 	1.48 10 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 OR 0.74 1.33 2.79	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2017) Menezes et al. (2017) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 6 2 2 2 2 32 64 6 2 32	679 guared 1500 2500 18 136 5200 78 438 1300 17200 guared 17200 guared 75 125 9 68 2000	=0.0187, p Cc Events 69 200 174 127 109 88 80 48 20 48 48 20 56 67 56 39	3004 =0.3046 C ontrol Total 104 268 200 200 182 200 200 182 144 120 1522 =0.1432 ntrol Total 52 152 152 134 100	2 0.1 0.2	0.5 1 2 Odds Ratio 	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 OR 0.74 1.33 2.79 0.97 1.02	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61] [0.57; 1.82]	100% W(fixed) 9.2% 18.7% 1.5% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7% 13.5%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Dionísio et al. (2017) Menezes et al. (2017) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Lanza et al. (2016)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 64 6 26 73	679 guared Total 150 250 18 136 520 78 438 130 1720 quared ental Total 75 125 9 68 2260 02	=0.0187, p Cc Events 69 200 174 127 109 88 80 48 48 48 48 48 48 48 48 48 48 48 48 48	3004 =0.3046 C Dentrol Total 104 304 200 200 182 200 182 200 182 182 144 120 1522 =0.1432 Total 52 152 134 100 100	2 0.2	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 0.99 0.98 1.38 1.43 1.60 1.18 OR 0.74 1.33 2.79 0.97 1.00 1.72	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.81; 2.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61] [0.51; 1.82] [0.67; 2.62]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7% 13.5% 20.5%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2017) Menezes et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2016) Jones et al. (2017)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 6 26 73 14	679 guared Total 150 250 18 136 520 78 438 130 1720 <i>quared</i> ental Total 75 125 9 68 260 39 9	=0.0187, p Cc Events 69 200 174 127 109 88 88 48 48 48 48 48 48 48 48 48 48 48	3004 =0.3046 C Dentrol Total 104 304 268 200 182 200 182 200 182 144 120 1522 =0.1432 Total 52 152 134 100 100 91	2 0.2	0.5 1 2 Odds Ratio 0.5 1 2 5 Odds Ratio	1.48 10 OR 0.68 1.34 0.99 0.98 1.38 1.43 1.60 1.18 OR 0.74 1.33 2.79 0.97 1.00 1.74	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61] [0.51; 1.82] [0.60; 1.68] [0.76; 3.95] [1.21, 2.72]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7% 13.5% 20.5% 8.2%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2016) Jones et al. (2017) Menezes et al. (2012)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 6 26 73 14 10 32	679 guared Total 1500 2500 18 136 5200 78 438 130 1720 quared 75 1255 9 68 2600 39 2200 65	=0.0187, p Cc Events 69 200 174 127 109 88 80 48 200 48 48 200 56 56 39 28 22 21	3004 =0.3046 C Dentrol Total 104 304 200 200 182 200 182 124 120 152 152 134 100 100 91 72 200	2 0.2	0.5 1 2 Odds Ratio 0.5 1 2 5 Odds Ratio	1.48 10 0R 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 0R 0.74 1.33 0.74 1.33 0.74 1.279 0.97 1.00 1.76 2.74 1.76	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61] [0.51; 1.82] [0.60; 1.68] [0.78; 3.95] [1.21; 3.79] [0.56 + 5.41]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7% 13.5% 20.5% 8.2% 16.5% 4.1%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Dionísio et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Dionísio et al. (2017) Menezes et al. (2017)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 6 26 73 14 103 9	679 guared Total 150 250 18 136 520 78 438 130 78 438 130 78 438 130 78 438 130 78 438 130 78 438 438 130 78 438 438 130 78 438 438 130 78 438 438 130 78 438 438 130 78 438 438 130 78 438 438 130 78 120 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 130 100 78 120 100 78 120 100 78 120 100 78 120 100 78 120 100 78 120 100 100 100 100 100 100 100 100 100	=0.0187, p Cc Events 69 200 174 127 109 88 80 48 20 48 48 20 56 39 28 22 21 5	3004 =0.3046 C Dentrol Total 104 304 200 182 200 182 124 120 1522 =0.1432 152 152 152 152 152 134 100 100 91 72 60	2 0.2	0.5 1 2 Odds Ratio 0.5 1 2 5 Odds Ratio	1.48 10 0R 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 0R 0.74 1.33 0.74 1.33 0.74 1.33 0.74 1.33 0.74 1.33 0.77 1.00 1.76 2.14 1.77	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61] [0.51; 1.82] [0.60; 1.68] [0.78; 3.95] [1.21; 3.79] [0.56; 5.61]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7% 13.5% 20.5% 8.2% 16.5% 4.1%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2016) Jones et al. (2016) Jones et al. (2017) Menezes et al. (2019) Fixed effect model	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 6 26 73 14 103 9	679 guared Total 150 250 18 136 520 78 438 130 78 438 130 78 438 130 78 438 130 78 438 130 78 438 438 130 78 438 438 130 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 520 78 438 438 520 78 438 438 520 78 438 438 520 78 438 438 520 78 438 520 78 438 438 520 78 520 78 438 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 520 78 53 520 78 53 520 78 53 520 78 53 520 78 53 520 78 53 520 78 53 520 520 53 520 520 53 53 53 53 53 53 53 53 53 53 53 53 53	=0.0187, p Cc Events 69 200 174 127 109 88 80 48 20 48 48 20 Events 26 67 56 39 28 22 21 5	3004 =0.3046 C Dentrol Total 104 304 200 182 200 182 120 182 121 144 120 1522 =0.1432 52 152 134 100 100 91 72 60 761	2 0.2	0.5 1 2 Odds Ratio	1.48 10 0R 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 0R 0.74 1.33 0.74 1.33 0.74 1.33 0.74 1.33 0.74 1.37 1.29	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61] [0.51; 1.82] [0.60; 1.68] [0.78; 3.95] [1.21; 3.79] [0.56; 5.61] [1.02; 1.63]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7% 13.5% 20.5% 8.2% 16.5% 4.1% 100%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2016) Jones et al. (2016) Jones et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=22	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 6 26 73 14 103 9	679 guared Total 150 250 18 136 520 78 438 130 78 125 9 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8	=0.0187, p Cc Events 69 200 174 127 109 88 80 48 200 48 48 20 Events 26 67 56 39 28 22 21 5	3004 =0.3046 C Dentrol Total 104 304 268 200 182 200 182 200 182 144 120 1522 =0.1432 52 152 134 100 100 91 72 60 761 .22549	2 0.2	0.5 1 2 Odds Ratio	1.48 10 0R 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 0R 0.74 1.33 0.74 1.279 0.97 1.00 1.76 2.14 1.77 1.29	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.81; 2.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61] [0.51; 1.82] [0.60; 1.68] [0.78; 3.95] [1.21; 3.79] [0.56; 5.61] [1.02; 1.63]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7% 13.5% 20.5% 8.2% 16.5% 4.1% 100%
Fixed effect model Heterogeneity: I-squared=14 (B) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Coelho et al. (2016) Jones et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=3 (C) Study Rizzo et al. (2003) Micheli et al. (2011) Andreeva et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2014) Saber-Ayad et al. (2016) Jones et al. (2016) Jones et al. (2017) Menezes et al. (2019) Fixed effect model Heterogeneity: I-squared=22 (D)	Experim Events 86 180 15 86 281 44 281 67 5.8%, tau-se Experim Events 32 64 6 26 73 14 103 9	679 guared Total 150 250 18 136 520 78 438 130 78 438 130 78 438 130 78 438 130 78 438 130 78 438 130 78 438 130 78 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 130 520 78 438 438 520 78 438 438 520 6 520 78 438 438 520 6 520 78 438 438 520 78 438 438 520 6 520 6 520 78 438 438 520 6 520 78 54 438 520 75 5 5 8 5 8 5 8 5 8 5 8 5 8 5 8 5 8 5	=0.0187, p Cc Events 69 200 174 127 109 88 88 48 48 48 48 48 48 48 48 48 48 48	3004 =0.3046 C Dentrol Total 104 304 268 200 182 200 182 144 120 1522 =0.1432 144 120 1522 134 100 100 91 72 60 761 .2549	2 0.2	0.5 1 2 Odds Ratio	1.48 10 OR 0.68 1.34 2.70 0.99 0.98 1.38 1.43 1.60 1.18 OR 0.74 1.33 2.79 0.97 1.00 1.76 2.14 1.77 1.29	[1.23; 1.77] 95%-Cl [0.41; 1.15] [0.93; 1.92] [0.76; 9.57] [0.63; 1.55] [0.71; 1.36] [0.98; 2.10] [0.97; 2.63] [1.01; 1.38] 95%-Cl [0.37; 1.51] [0.83; 2.14] [0.67; 11.61] [0.51; 1.82] [0.60; 1.68] [0.78; 3.95] [1.21; 3.79] [0.56; 5.61] [1.02; 1.63]	100% W(fixed) 9.2% 18.7% 1.5% 12.1% 23.1% 8.7% 16.9% 9.8% 100% W(fixed) 10.7% 24.0% 2.7% 13.5% 20.5% 8.2% 16.5% 4.1% 100%

Figure 2. Forest plots of significant PRO and NP comparison models. (A) *ACTN3* R vs. X allele in PROs; (B) *ACTN3* RR vs. RX+XX in PROs; (C) *ACE* D vs. I allele in NPs; (D) *ACE* DD vs. DI+II in NPs.