Genome-wide linkage scan for athlete status in 700 British female DZ twin pairs


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ssociation studies, comparing elite athletes with sedentary controls, have reported a number of genes that may be related to athlete status. The present study reports the first genome-wide linkage scan for athlete status. Subjects were 4488 adult female twins from the TwinsUK Adult Twin Registry (793 monozygotic [MZ] and 1000 dizygotic [DZ] complete twin pairs, and single twins). Athlete status was measured by asking the twins whether they had ever competed in sports and what was the highest level obtained. Twins who had competed at the county or national level were considered elite athletes. Using structural equation modeling in Mx, the heritability of athlete status was estimated at 66%. Seven hundred DZ twin pairs that were successfully genotyped for 1946 markers (736 microsatellites and 1210 SNPs) were included in the linkage analysis. Identical-by-descent probabilities were estimated in Merlin for a 1 cM grid, taking into account the linkage disequilibrium of correlated SNPs. The linkage scan was carried out in Mx using the \( \pi \)-approach. Suggestive linkages were found on chromosomes 3q22-q24 and 4q31-q34. Both areas converge with findings from previous studies using exercise phenotypes. The peak on 3q22-q24 was found at the SLC9A9 gene. The region 4q31-q34 overlaps with the region for which suggestive linkages were found in two previous linkage studies for physical fitness (FABP2 gene; Bouchard et al., 2000) and physical activity (UCP1 gene; Simonen et al., 2003). Future association studies should further clarify the possible role of these genes in athlete status.

Elite athletic performance is thought to be the result of the combination of a high genetic potential and the optimal environmental conditions, such as training and nutrition (MacArthur & North, 2005). It has long been recognized that physical performance is genetically determined (Bouchard & Malina, 1998), as evidenced by the significant heritabilities ranging from 31% to 85% for physical performance traits involving different aspects of cardiorespiratory fitness (Bouchard et al., 1998; Bouchard et al., 1999; Perusse et al., 2001) and skeletomuscular strength and performance (De Mars et al., 2007; Thomis et al., 1998).

The number of studies aiming to identify the actual genetic variants that account for the heritability of physical performance phenotypes is increasing and are summarized in the 'human gene map for performance and health-related exercise phenotypes', which is regularly updated (Perusse et al., 2003; Rankinen et al., 2001; Rankinen et al., 2002; Rankinen et al., 2004; Wolfarth et al., 2005). Besides linkage and association studies on different physiological parameters related to exercise, the human gene map also includes a number of association studies of candidate genes, in which elite athletes are compared with sedentary controls. Genes that have been related to elite athlete status are, for example, the angiotensin-converting enzyme (ACE) gene (Woods et al., 2000), and the bradykinin receptor (B2R) gene (Williams et al., 2004). These genes are thought to have an impact on physical performance through increasing cardiorespiratory fitness as a result of training, but possibly also through enhanced muscle efficiency (Williams et al., 2000). Other genes that have been related to elite athlete status are the \( \alpha \)-actin-3 (ACTN3) gene (MacArthur & North, 2007; Paparini et al., 2007; Yang et al., 2003) and two adrenergic receptor genes (ADRA2A and ADRB2; Moore et al., 2001; Wolfarth et al., 2000).

To date, no linkage studies have been published that identified the genomic locations that may be related to athlete status. In this study, the results of a heritability analysis and genome-wide linkage scan are presented for data on athlete status in a sample of British female twin pairs. Athlete status was measured by asking the twins whether they ever participated in sports and what was the highest
level they ever competed at. Twins who competed at the national or county level were considered elite athletes.

**Methods**

**Sample**

Phenotype information came from a mailed survey sent out in 2000 to female twins registered at the TwinsUK Adult Twin Registry (Spector & Williams, 2006). There were 4527 twins with valid data on the level of participation in competitive sports. Twins with unknown zygosity were excluded (39 individuals), resulting in a total sample of 4488 individuals, from which 1793 complete twin pairs could be formed (i.e., both individuals had data on sports level). Table 1 gives an overview of the number of complete and incomplete twin pairs as a function of zygosity. Zygosity of the twin pairs was determined using a standardized questionnaire and DNA fingerprinting (Spector et al., 1996). The mean age of the individuals was 51.9 (standard deviation: 12.8), the minimum age was 20-years-old and the maximum 83-years-old.

**Phenotyping**

Athlete status was measured by asking the participants to indicate for a list of sports whether they had ever participated in each of these sports and what was the highest level at which they had ever competed in these sports. The levels were social only, in a school, club or university team, at the county level or at the national level. Table 2 lists the sports that were included in the survey. Yoga and walking were not included in the analyses, since these were low-impact sports (at less than 4 metabolic equivalents). Participants were classified into three categories: (1) never participated in sports in any organized form (no exercise and exercise at the social level), (2) ever participated in sports at the school, club or university level, and (3) ever participated in sports at the county or national level. Women who ever participated in sports at the county or national level are thought to have been elite athletes (MacArthur & North, 2005).

**Genotyping**

DNA of 2600 individuals was extracted from venous blood samples following standard protocols. Two different marker sets were used: a microsatellite marker set using standard ABI Prism genotyping methodologies (Applied Biosystems, Foster City, CA, USA) and a single nucleotide polymorphism (SNP) marker set (HuSNP GeneChips, Affymetrix), which are more fully described elsewhere (Wilson et al., 2003). Basic checks for sex and zygosity were carried out. Errors of Mendelian inheritance were detected using Pedstats (Abecasis et al., 2002). Unlikely recombinants were detected using Merlin (Abecasis et al., 2002). Unlikely or erroneous genotypes due to Mendelian or genotyping errors were set as unknown. SNPs with minor allele frequency lower than .05 were excluded. Genetic map positions of the markers were taken from the interpolated genetic map developed by Duffy (2006). For markers that were not present on this map, their physical position was taken from the National Center for Biotechnology Information (NCBI) build 35.1. Locally weighted linear regression of the genetic map positions on the physical positions was used to impute the missing genetic positions. Markers that could not be reliably located on the physical map (which was the case for some SNPs) were excluded from the dataset. This resulted in 1946 markers, of which 736 were microsatellites and 1210 SNPs. All genotyped dizygotic (DZ) twin pairs with complete data on athlete status were selected (705 pairs). Twin pairs for whom one of the individuals had less than 200 successfully genotyped markers were excluded (5 pairs). These were pairs with only microsatellites genotyped. There were 99 pairs with less than 200 successfully genotyped microsatellites, but with a large number of SNPs. These pairs were retained in the analyses. Thus, the final sample used for the linkage analysis consisted of 700 pairs. The mean age of these individuals was 54.6 (standard deviation: 11.5), the minimum age was 21 and the maximum 83 years old. Average spacing of all 1946 markers was 6.22 centiMorgan (cM) and the average heterozygosity 61.0%. For the statistical analyses the Haldane mapping function was used. All reported values are in Haldane cM.

Markers that were close together on the genetic map, as was the case for some of the SNPs in the dataset, may be in linkage disequilibrium. If linkage disequilibrium among markers is ignored in the linkage analysis, this might lead to an upward bias in linkage signals (Abecasis & Wigginton, 2005). Merlin accommodates a feature that identifies clusters of markers that are in linkage disequilibrium, and estimates the haplotype frequencies for the clustered markers. Using an r-squared of .30 as a cut-off value, clusters of markers and their haplotype frequencies were estimated. This resulted in 73 clusters, of which 62 contained two markers, 10 contained 3 markers, and 1 contained 4 markers. Three clusters contained both SNPs and microsatellites. For markers that were not clustered, the allele frequencies were determined by counting the alleles that occurred in the dataset. The obtained allele and

### Table 1

<table>
<thead>
<tr>
<th></th>
<th>Complete twin pairs</th>
<th>Incomplete twin pairs</th>
<th>Total pairs (individuals)</th>
<th>Polychoric twin correlation (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>MZ</td>
<td>793</td>
<td>348</td>
<td>1141 (11934)</td>
<td>.66 (0.59–0.71)</td>
</tr>
<tr>
<td>DZ</td>
<td>1000</td>
<td>554</td>
<td>1554 (2554)</td>
<td>.32 (0.24–0.40)</td>
</tr>
<tr>
<td>Total</td>
<td>1793</td>
<td>902</td>
<td>2695 (4488)</td>
<td></td>
</tr>
</tbody>
</table>

Note: 95% CI = 95% confidence interval; MZ = monozygotic; DZ = dizygotic
haplotype frequencies were used to estimate the identical-by-descent (IBD) probabilities, thereby assuming that within clusters the recombination fraction is zero and across clusters there is linkage equilibrium (Abecasis & Wigginton, 2005).

### IBD Estimation

A sibling pair shares an allele at a specific genomic locus identical by descent (IBD) if the allele is inherited from the same parent. Thus, a sibling pair can share 0, 1, or 2 alleles IBD at a specific locus. If parental genotypes are not available, the probabilities of IBD status 0, 1, or 2 were estimated based on the allele frequencies as observed in the population. The IBD status is evaluated by the log-likelihood ratio test (LRT). The difference in minus two times the log-likelihood 

\[ \Delta \chi^2 \] 

between the two models. If the \( \Delta \chi^2 \)-test yielded a \( p \)-value higher than .05, the fit of the submodel was not significantly worse than the fit of the full model, and the submodel was kept as the most parsimonious and best fitting model.

In the sample of genotyped DZ twin pairs, the DZ twin correlation and the thresholds were computed to examine whether the genotyped sample deviated from the total sample with respect to the DZ twin correlation and the prevalence of sports participation at the various levels. Next, an AQE model was fitted to the data to test, at each locus, whether the proportion of alleles shared IBD in each DZ twin pair explained their resemblance in athlete status (Neale et al., 2003). Using the previously given formula, \( \hat{\pi} \) was computed in Mx from the estimated IBD probabilities. At each locus, the

### Table 2

<table>
<thead>
<tr>
<th>No sports</th>
<th>School/Club/University level</th>
<th>County/National level</th>
</tr>
</thead>
<tbody>
<tr>
<td>N (%)</td>
<td>N (%)</td>
<td>N (%)</td>
</tr>
<tr>
<td>1. Swimming</td>
<td>3849 (85.8)</td>
<td>567 (12.6)</td>
</tr>
<tr>
<td>2. Cycling</td>
<td>4463 (99.4)</td>
<td>23 (0.5)</td>
</tr>
<tr>
<td>3. Running</td>
<td>3656 (81.5)</td>
<td>742 (16.5)</td>
</tr>
<tr>
<td>4. Keep fit/Aerobics</td>
<td>4382 (97.6)</td>
<td>101 (2.3)</td>
</tr>
<tr>
<td>5. Gymnastics</td>
<td>4092 (91.2)</td>
<td>377 (8.4)</td>
</tr>
<tr>
<td>6. Tennis</td>
<td>3983 (88.7)</td>
<td>487 (10.9)</td>
</tr>
<tr>
<td>7. Badminton</td>
<td>4195 (93.5)</td>
<td>280 (6.2)</td>
</tr>
<tr>
<td>8. Squash</td>
<td>4413 (98.3)</td>
<td>68 (1.5)</td>
</tr>
<tr>
<td>9. Golf</td>
<td>4437 (98.9)</td>
<td>46 (1.0)</td>
</tr>
<tr>
<td>10. Skiing</td>
<td>4468 (99.6)</td>
<td>18 (0.4)</td>
</tr>
<tr>
<td>11. Ice skating</td>
<td>4480 (99.8)</td>
<td>8 (0.2)</td>
</tr>
<tr>
<td>12. Yoga</td>
<td>4472 (99.6)</td>
<td>13 (0.3)</td>
</tr>
<tr>
<td>13. Dancing</td>
<td>4351 (96.9)</td>
<td>122 (2.7)</td>
</tr>
<tr>
<td>14. Rugby/Football</td>
<td>4423 (98.6)</td>
<td>61 (1.4)</td>
</tr>
<tr>
<td>15. Cricket</td>
<td>4433 (98.8)</td>
<td>52 (1.2)</td>
</tr>
<tr>
<td>16. Martial arts</td>
<td>4457 (99.3)</td>
<td>26 (0.6)</td>
</tr>
<tr>
<td>17. Boxing</td>
<td>4485 (99.9)</td>
<td>2 (0.0)</td>
</tr>
<tr>
<td>18. Hill walking</td>
<td>4463 (99.4)</td>
<td>24 (0.5)</td>
</tr>
<tr>
<td>19. Walking</td>
<td>4444 (99.0)</td>
<td>40 (0.9)</td>
</tr>
<tr>
<td>20. Other</td>
<td>3765 (83.9)</td>
<td>597 (13.3)</td>
</tr>
<tr>
<td>Any sport</td>
<td>2498 (55.7)</td>
<td>697 (37.4)</td>
</tr>
</tbody>
</table>

Note: \( N = \) number of women; \( \% = \) percentage of women.

The numbers given for the individual sports do not add up to the numbers given for any sport, because some women had participated in more than one sport.
significance of the effect of Q was evaluated by use of the likelihood ratio test. From this test, the LOD score can be computed by dividing the obtained chi-square test statistic by 2ln10 (~ 4.6).

Empirical thresholds for suggestive and significant linkage were computed by randomly permuting the datasets 1000 times. Permutations were carried out by randomly assigning the IBD estimates to the sibling pairs, keeping the sibling pairs and IBD structure of the whole genome intact. A linkage scan was then performed on each permuted dataset. The empirical threshold for suggestive linkage was computed by obtaining the maximum LOD score for each chromosome out of the 1000 analyses, and determining what LOD score occurred 1000 times out of 22,000. The threshold for significant linkage was computed by recording the maximum LOD score in each linkage scan on each permuted dataset, and then determining which LOD score occurred 50 out of 1000 times (see also Boomsma et al., 2006). The empirical threshold for suggestive linkage was 1.63 and for significant linkage 2.96.

## Results

### Prevalence of Sports Participation at the Various Levels

From 4488 women, there were 311 women who ever participated in sports at the county or national level (6.9%). Sixteen hundred and seventy-nine women ever participated in sports at the school, club, or university level (37.4%). The remaining 2498 women never participated in any organized form of sports (55.7%). There were no differences in these prevalences between first and second-born twins ($\chi^2 = 3.53, p = .47$) nor between MZ and DZ twins ($\chi^2 = 2.50, p = .64$). Women from older birth cohorts less frequently participated in sports at the school, club, or university level, and the county or national level, as indicated by the significant and positive regression coefficients of age on both thresholds ($\chi^2 = 100.64, p < .001$) and the second threshold ($\chi^2 = 33.63, p < .001$).

### Heritability of Athlete Status

The polychoric MZ twin correlation was .66 (95% Confidence Interval (CI): 0.59–0.71) and the polychoric DZ twin correlation .32 (95% CI: 0.24–0.40). We fitted an ADE model to the data. The thresholds were modeled according to the best fitting constrained saturated model, which contained two thresholds and a cohort effect on each threshold. The model fitting results of the heritability analyses of athlete status are given in Table 3. The AE-model described the data adequately. Of the total variance in liability for athlete status, 65.5% was explained by A (95% CI: 59.6%–70.6%) and 34.5% by E (95% CI: 29.3%–40.4%). Thus, a combination of additive genetic and nonshared environmental factors influences athlete status.

### Genome-Wide Linkage Scan for Athlete Status

The prevalence of sports participation at the various levels in the genotyped DZ twins was highly similar to the percentages in the total sample: 6.8% ever participated in sports at the county or national level, 36.9% ever participated at the school, club, or university level, and 56.3% did not ever participate in sports at the organized level. The twin correlation in the genotyped DZ twin pairs was .33. Thus, the genotyped sample was representative for the total sample in terms of both prevalence and the DZ twin correlation.

The results of the genome-wide linkage scan are depicted in Figure 1. Suggestive linkages were found on chromosomes 3q24 and 4q32.3. The peak on 3q24 has a maximum LOD score of 2.35 at 153 cM. The confidence interval of this peak, using the drop 1 LOD rule, is 3q22–q24 (145–160 cM). The markers at the peak are D3S1569 and the clustered markers rs2801 and rs2800, all residing within the sodium/hydrogen exchanger 9 (SLC9A9) gene. The second peak is on 4q32.3, has a maximum LOD score of 1.87, and is found at 168/169 cM near marker D4S1597. The confidence interval of the peak is 4q31–q34 (147–188 cM).

### Table 3

Model Fitting Results for Heritability Analyses of Athlete Status

<table>
<thead>
<tr>
<th>Model</th>
<th>(-2LL)</th>
<th>(df)</th>
<th>Vs.</th>
<th>$\chi^2$</th>
<th>$\Delta df$</th>
<th>$p$</th>
<th>A</th>
<th>D</th>
<th>E</th>
<th>$r_1$</th>
<th>$r_2$</th>
<th>$\beta_1$</th>
<th>$\beta_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1. Saturated</td>
<td>7479.069</td>
<td>4476</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2. ADE</td>
<td>7483.820</td>
<td>4482</td>
<td>1</td>
<td>4.751</td>
<td>6</td>
<td>.576</td>
<td>62.5</td>
<td>0.03</td>
<td>34.4</td>
<td>-0.720</td>
<td>0.797</td>
<td>0.017</td>
<td>0.014</td>
</tr>
<tr>
<td>3. AE</td>
<td>7483.852</td>
<td>4483</td>
<td>1</td>
<td>4.783</td>
<td>7</td>
<td>.886</td>
<td>65.5</td>
<td></td>
<td>34.5</td>
<td>-0.719</td>
<td>0.797</td>
<td>0.017</td>
<td>0.014</td>
</tr>
<tr>
<td>4. E</td>
<td>7767.048</td>
<td>4484</td>
<td>3</td>
<td>283.196</td>
<td>1</td>
<td>&lt; .001</td>
<td>100</td>
<td>-0.673</td>
<td>0.799</td>
<td>0.016</td>
<td>0.014</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Note: \(-2LL = -2\) log-likelihood; \(df = \) degrees of freedom; vs. = model to which the fitted model is compared; $\chi^2 = \) chi-square value; $\Delta df = \) difference in degrees of freedom between fitted and comparison model; $p = \) p-value; A = proportion of variance explained by additive genetic factors; D = proportion of variance explained by non-additive genetic factors; E = proportion of variance explained by nonshared environmental factors; $r_1 = \) first thresholds; $r_2 = \) second threshold; $\beta_1 = \) regression coefficient of age on the first threshold; $\beta_2 = \) regression coefficient of age on the second threshold; Positive age regression coefficients indicate that athlete status was less frequent in older women. The most parsimonious and best fitting model is shown in bold.
Figure 1
LOD scores of the genome-wide linkage scan for athlete status in the sample of genotyped DZ twin pairs.

Note: dashed line indicates the empirical threshold for suggestive linkage.
Discussion

Seven per cent of the British adult women have ever competed in sports at the county or national level. Women from older cohorts have less frequently participated in sports in general and also at the county or national level. Athlete status in women is a heritable trait: around 66% of the variance in athlete status is explained by additive genetic factors. The remaining variance is due to nonshared environmental factors. Nonadditive genetic factors and shared environmental factors do not play a role in explaining individual differences in athlete status.

The heritability of athlete status is consistent with the significant heritabilities that have been reported for a range of measures of athletic abilities. For example, the heritabilities of different measures of oxygen uptake, such as maximal oxygen uptake in the sedentary state and in response to training and submaximal oxygen uptake at different power outputs, which are all indicators of aerobic capacity, ranged from 47% to 74% (Bouchard et al., 1998; Bouchard et al., 1999). Heritabilities of muscle strength phenotypes (muscle cross-sectional area and isometric, concentric, and eccentric strength) ranged from 31% to 85% (Thomis et al., 1998), with the majority of the genetic variance shared among the strength phenotypes (De Mars et al., 2007).

Suggestive linkage for athlete status was found on chromosome 3q24 (LOD = 2.35) at the SLC9A9 gene. This gene encodes a cell membrane transport protein that is part of the sodium/hydrogen exchange family and is expressed at high levels in heart and skeletal muscle. SLC9A9 has been related to attention deficit hyperactivity disorder (De Silva et al., 2003), but not to sports ability. We also examined whether there were genes listed in the ‘human gene map’ that were located at or nearby the peak on 3q24 (Wolfarth et al., 2005). The calcium-sensing receptor (CASR) gene is located just outside the confidence interval of the peak on 3q21. The CASR gene has been found to be associated with physical activity (hours per week during the last year) in adolescent girls, and is involved in calcium homeostasis (Lorentzon et al., 2001). It is well known that calcium is involved in muscle activity, bone formation, blood clotting, and nerve activity (Wilmore & Costill, 2004).

Suggestive linkage was also found on chromosome 4q32 (LOD = 1.87). There are two genes that are located in the region 4q28-31 that are listed in the ‘human gene map’. The first is the fatty-acid binding protein 2 (FABP2) gene. In a linkage study carried out in 412 sibling pairs using 289 markers (Bouchard et al., 2000), suggestive linkage ($p < .01$) was found on 4q26 at marker FABP2 for maximal oxygen uptake in response to a 20 week exercise training program. Fatty-acid binding proteins are involved in the metabolism of long-chain fatty-acids. Through its involvement in metabolic processes, one could hypothesize that the FABP2 gene is related to sports performance and participation. The second gene located in the 4q28-31 region that is listed in the ‘human gene map’ is the uncoupling protein 1 (UCP1) gene. The UCP1 gene is involved in heat generation. In a sample of 395 offspring and 372 parents using 432 markers, suggestive linkage ($p < .01$) was found at the UCP1 gene for moderate to strenuous daily physical activity (Simonen et al., 2003). To conclude, the peaks on 3q and 4q reported in the present study are of potential interest because they are close to genes that have been previously related to aerobic fitness and physical activity.

A first limitation of this study is that we used a very general measure of athlete status. Because the grouping also reflected the contrast between regular exercisers and sedentary subjects, athlete status was partly confounded with voluntary exercise participation. Substantial heritability for leisure-time exercise participation (Simonen et al., 2002; Stubbe et al., 2006), has been found, ranging from 16 to 71%. Genes for athletic status may well constitute part of the heredity of voluntary regular exercise behaviour. Individuals with the genetic advantage to perform well at sports are more likely to continue participating in sports. They find that they are good at sports and they possibly also experience the physical need to be active, which together drives them to regularly participate in exercise. A G x E scenario may unfold where regular participation in exercise by those individuals ‘who are good at it’ provides the additional training that is needed to maximize actual performance, ultimately leading to elite athlete status (Brutsaert & Parra, 2006).

A second limitation is that the sports for which the twins could indicate their maximum level of competition were biased in favor of endurance sports, requiring mainly aerobic fitness, rather than strength or sprint capacity. The genes affecting endurance or power athlete status might be different, or different alleles of the same gene might be associated with endurance or power athlete status, the latter having been suggested for the ACE gene (Williams et al., 2000), and more recently also for the ACTN3 gene (MacArthur et al., 2007). It is increasingly acknowledged however that the distinction between endurance and power sports is not that sharp; many endurance sports also contain power elements and vice versa (Wilmore & Costill, 2004).

A third limitation of the study involves the inclusion of women from different ages in the sample. One could argue that the younger women (less than 30 years) might not have reached their maximum level of competition yet, which could have affected the linkage results. Still, the prevalence of competition at the county or national level is highest in the...
younger adults, also in women aging between 20 and 25 years. The prevalence of competition at the county or national level is smallest in the oldest cohorts. The significant cohort effect on the prevalence of sporting ability can be explained by the fact that in the past it was less common for women to participate in sports at a competitive level. This cohort effect is unlikely to have biased the results of the genetic analyses however, since we regressed out the cohort effect in the analyses. It is possible, though, that the heritability of athlete status is higher in older cohorts since it was rarer, but we do not have the power to detect this.

A final limitation of this study relates to the sample size for the linkage scan. It is well known that in order to detect linkage signals, large sample sizes are needed, especially when categorical phenotypes are analysed (Neale et al., 1994). We therefore investigated the power to detect QTL effects of different magnitudes, given a sample size of 700 sibling pairs and a total heritability of 65%. The power to detect a QTL effect of 30% of the total variance is 0.91, and the power to detect a QTL effect of 25% of the variance is 0.79. For QTL effects explaining, respectively, 20%, 15%, and 10% of the variance, the power is 0.60, 0.38, and 0.20. Thus, the power to detect linkage signals of 25% or larger is good, and the power to detect smaller effects is moderate to low. The limited power of the present study underlines the need for replication studies.

To summarize, it was shown that athlete status in women is a heritable trait. A genome-wide linkage scan revealed suggestive linkages on chromosomes 3q22-q24 and 4q31-q34. The peak at 3q22-q24 is found at the SLC9A9 gene. The region 4q31-q34 overlaps with the region for which suggestive linkages were found in two previous linkage studies for physical fitness (FABP2 gene; Bouchard et al., 2000) and physical activity (UCP1 gene; Simonen et al., 2003). Future association studies should further clarify the possible role of these genes in athlete status.

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References


