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A Genome-wide association study of self-rated health

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Abstract

Self-rated health questions have been proven to be a highly reliable and valid measure of overall health as measured by other indicators in many population groups. It also has been shown to be a very good predictor of mortality, chronic or severe diseases, and the need for services, and is positively correlated with clinical assessments. Genetic factors have been estimated to account for 25 - 64% of the variance in the liability of self-rated health. The aim of the present study was to identify Single Nucleotide Polymorphisms (SNPs) underlying the heritability of self-rated health by conducting a genome-wide association analysis in a large sample of 6,706 Australian individuals aged 18–92. No genome wide significant SNPs associated with self-rated health could be identified, indicating that self-rated health may be influenced by a large number of SNPs with very small effect size. A very large sample will be needed to identify these SNPs.

Keywords

genome-wide association; genes; self-rated health; self-reported health; health

Introduction

Self-rated health questions have been developed with the objective of quantifying an individual's perception of his or her overall health state. Even though these measures tend to be less sensitive to changes in specific disorders (Beaton & Schemitsch, 2003), it have been proven to be a highly reliable and valid measure of overall health as measured by means of other indicators in different population groups (Lundberg & Manderbacka, 1996). It has also been shown to be a very good predictor of mortality and the need for services (Leinonen, et al., 2005), and is positively correlated with clinical assessments (Romeis, et al., 2000). Furthermore, higher self-rated health has been associated with absence of chronic diseases, severe diseases, disabilities, functional limitations, and with higher physical activity, and better psychosocial wellbeing (Bryant, Beck, & Fairclough, 2000; Idler, 1993; Leinonen, Heikkinen, & Jylha, 2001; Rodin & McAvay, 1992). Typically, self-rated health is based on a single question asking the respondents to rate their current health status. Most individuals rate their health as moderate to good while few would rate their health as bad (Juerges, 2007).

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Several twin studies have investigated the heritability of self-rated health (Christensen, Holm, McGue, Corder, & Vaupel, 1999; Harris, Pedersen, McClearn, Plomin, & Nesselroade, 1992; Leinonen, et al., 2005; Lichtenstein & Pedersen, 1995; Romeis, et al., 2000; Silventoinen, Posthuma, Lahelma, Rose, & Kaprio, 2007; Svedberg, Lichtenstein, & Pedersen, 2001) estimating genetic factors to account for 25 - 64 % of the variance in the liability of self-rated somatic health. A large longitudinal study of Finnish twins showed that the heritability of self-rated health was greatest at age 16, at 63% (95% CI: 0.56 - 0.67), declining steadily to age 25 with a heritability of 33% (CI: 0.25 - 0.41) (Silventoinen et al., 2006). The study revealed moderate correlations between the different health ratings at different life stages (r = 0.33-0.61), which were predominately due to genetic factors. The finding of decreasing heritability of self-rated health with age, however, is not confirmed by cross-sectional studies, for example Mosing et al. (2009) found a heritability of 46% in an elderly twin sample (mean age = 61 ± 8.8). As self-rated health has been shown to be for a substantial part due to genetic factors, it would be interesting to explore the genetic variants underlying this trait. The aim of the present study was to identify SNPs underlying the heritability of self-rated health by conducting a genome-wide association (GWA) analysis on a large sample of Australian individuals who have previously rated their health status.

Methods

Participants and measures

Self-rated health data were collected in four twin family studies conducted at the Queensland Institute of Medical Research (QIMR). The two earliest studies (Study 1 and Study 2) were conducted between 1993 and 1995. Study 3 was conducted between 1996 and 2000 and the most recent and largest study (Study 4), was collected between 2001 and 2005 (see Table 1a for more details). All four studies consisted of mailed-out questionnaires assessing health and lifestyle issues as well as demographic information and were approved by the QIMR Human Research Ethics Committee. Content and sampling methods of the four studies have been described in detail elsewhere (Bucholz, et al., 1998;Hansell, et al., 2008;Heath, et al., 1997;Mosing, Gordon, et al., 2009;Mosing, Zietsch, et al., 2009). In Study 3 and 4 (more than 95% of the final sample), self-rated health was assessed with the following item: "How would you describe your general physical health?", rated on a 4-point Likert scale, "excellent" (1), "good" (2), "fair" (3), "poor" (4). In the other two studies the self-rated health questions were worded slightly differently: "In general, would you say that your physical health now is excellent, good, fair or poor?" (Study 1) and "How would you describe your health at present? - very good, good, fair, poor, very poor" (Study 2). As few individuals rated their health as poor or very poor and for consistency with the self-rated health questions of the other studies, which only had four instead of five categories, the categories poor and very poor of the self-rated health item in Study 2 were collapsed.

Self-rated health and genotype data were available for 6,706 individuals (3,710 females and 2,996 males) from 2,585 independent families aged between 18–92 years (Mean=46; SD=11). Of these, 1,403 (21%) individuals participated in more than one of the studies in which case the most recent rating was used. Test-retest Pearson correlations between the different self-rated health measures ranged between 0.48 and 0.65 and the correlation between the two identically worded items was not higher compared to the questions used in the other two studies. Table 1a shows the number of individuals derived from each study forming the final sample. Finally, in line with previous findings, most individuals rated their health as good while few reported poor health, resulting in a skewed distribution (Table 1b). Therefore, a square root transformation was applied to the final scores and the scores were treated as continuous. Previous behaviour genetic analysis in a subsample of the present sample revealed heritability estimates of 46% with the remaining variance being due to non-shared environmental influences (Mosing, Zietsch, et al., 2009).

Genotyping, quality control, and imputation procedures

Over more than twenty years a wide range of phenotypic data and DNA samples have been collected as part of the different projects. The DNA samples were collected in accordance with standard protocols and were genotyped using the following Illumina SNP platforms: 317K, HumanCNV370-Quadv3, and Human610-Quad. Quality control (QC) procedures employed are discussed in full detail in Medland et al. (2009). Briefly, checks for ancestry outliers, Mendelian errors, Hardy Weinberg Equilibrium, and Minor Allele Frequency (MAF) were conducted separately for each of the projects and then again for the combined dataset. The final dataset consisted of 269,840 SNPs and was imputed by MACH (Abecasis, unpublished) using the data from the European HapMap 1+2, Release 22 Build 36. SNPs with an imputation quality score (r²) greater than 0.3 were retained resulting in a total of 2,380,486 imputed SNPs. Finally, if only one individual from a monozygotic twin pair had been genotyped, the non-genotyped co-twin was assigned that genotype as well.

Statistical analyses

The best guess genotype at each SNP was tested for association with self-rated health using the family-based association test in Merlin (Chen & Abecasis, 2007) accounting for family relationships. The additive genetic effect was computed by modelling the genotypic mean of the heterozygote (Aa) as the average of the two homozygotes (AA, aa). The generally accepted genome-wide significance level for the association between SNP and phenotype at $\alpha = 0.05$ is $7.2*10^{-8}$ or smaller, correcting for the total number of independent tests (Dudbridge & Gusnanto, 2008), and was also applied in the present study.

Additionally, a gene-based test (VEGAS), feasible for use with GWA data with related individuals (Liu, et al., accepted), was conducted to test whether there are any genes which harbour an excess of associated variants. Details of this procedure are summarized elsewhere (Liu, et al., accepted; Verweij, et al., accepted). In brief, this test explores association on a per-gene basis taking the p-values of all SNPs within 50kb of each gene, as well as linkage disequilibrium (LD) and number of SNPs per gene into account. A p-value below $\alpha = 2.8 \times 10^{-6}$ was considered to be significant as the gene-based association test included 17,585 genes (0.05/17,585).

Finally, power calculations were conducted by performing association tests on simulated datasets (based on our sample) in Merlin. The simulated datasets maintain the features of the original data in terms of marker informativeness, allele frequency, spacing, missing data patterns, and trait distribution despite replacing the phenotypic values and the genotypes for a randomly selected SNP with a minor allele frequency of 0.25. One-thousand simulated data sets were generated on which association analyses were subsequently performed. Detailed information on the simulation procedure can be found on http://www.sph.umich.edu/csg/abecasis/Merlin/reference/simulation.html. The empirical power estimate is given by the proportion of genome-wide significant association tests detected in the 1000 association analyses. The present sample provides 99% and 50% power to detect SNPs explaining 1% and 0.5% of variance in self-rated health, respectively.

Results

We tested 2,380,486 SNPs for association with self-rated health correcting for age and sex. The Quantile-Quantile (QQ) plot (Figure 1) shows the association between the observed versus the expected (under the null-hypothesis of no association) p-values of the autosomal associations.

Results of the association analysis (shown in Figure 2) indicate that there are no genomewide significant ($\alpha = 7.2*10^{-8}$) association signals, with the smallest p-value of $2.3*10^{-7}$

obtained for a SNP (rs17043947) on chromosome 2p24.1. However, though not significant, we found two promising regions on chromosome two (2p24.1 and 2q14.3) with the top hits having p-values of $2.3*10^{-7}$ and $7.6*10^{-7}$, respectively. Table 2 shows SNPs in the top 50 smallest p-values for self-rated health. Redundant SNPs in high LD ($r^2 > .70$) with a more significant SNP were excluded.

The gene-based test did not reveal significant results ($\alpha = 2.8 \times 10^{-6}$), with the smallest p-value being 9.0*10⁻⁵. Table 3 shows the five genes with the smallest p-values.

Discussion

The present study is the first to perform a genome-wide association analysis on self-rated health. Despite the high power (99%) to detect SNPs accounting for 1% of the variance in self-rated health, no genome-wide significant SNPs were identified. However, though not significant, we found two promising regions on chromosome two. Also, some of the top 50 SNPs were close to genes (e.g. *MAML2, PROM1, PROC*) broadly associated with a variety of health conditions, such as inflammation, coronary disease, cardio vascular disease, thrombosis, protein C deficiency etc (Reiner, et al., 2008; Trynka, et al., 2009; Wu, et al., 2009). The Protein C (*PROC*) gene was also in the top 5 genes revealed by the gene-based test. Changes in these genes may have an effect on an individuals' self-rating of health. As no other study has explored the molecular genetic basis of self-rated health we cannot compare our findings.

Nevertheless, the fact that we did not find a genome-wide significant SNP is not totally unexpected considering that self-rated health is a very broad measure, influenced not only by several general somatic health factors but also by the health status presently experienced at the time of the rating. The concept of self-rated health has also been shown to be strongly associated with mental health, e.g. someone who is depressed may rate their health status as lower than someone who is in a good state of mind. Additionally, particular personality traits may play a role in how an individual rates his or her own health, for example a person scoring high in neuroticism would most likely rate their health slightly worse than a person very low in neuroticism. All these facts indicate that self-rated health is a very broad concept on a phenotypic level and may be genetically even more complicated. We suggest that very many rare variants of small effect size may influence self-rated health and are therefore difficult to detect. The fact that our Q-Q plot (Figure 1) lifts appreciable above the 95% confidence interval also hints at the highly polygenic nature of our trait, self-rated health. A recent paper by Yang et al. (2010) showed that 45% of the variance of human height could be explained considering all SNPs in a study (294,831), as opposed to 5% explained by SNPs detected by the conventional GWAS approach. This indicates that even in a very clear-cut and highly heritable phenotype such as human height, variance is explained by a large number of SNPs with very small effect; too small to be detected in a normal GWAS. The International Schizophrenia Consortium showed that by using the top-half (p-value below 0.5) of the SNPs, they could quite consistently predict Schizophrenia and related disorders (e.g. Bipolar disorder) in other samples, supporting the idea of a polygenic basis to the phenotype (Purcell, et al., 2009). Another study on human height by Lango Allen et al. (submitted) also supports these findings: with a sample of almost 200,000 individuals they show that hundreds of genetic variants influence variance in adult height. This also shows that in order to find genes with such a small effect size a very large sample is needed. Aiming for this, a large consortium has been founded planning to conduct a meta-analysis on self-rated health in the near future, combining several samples in order to possibly confirm the regions of interest found in the present study and find additional genetic variants underlying the variation of self-rated health and possibly even predict self-rated health and related health measures/indicators across different samples as in Purcell et al. (2009).

In summary, no genome wide significant SNPs underlying self-rated health could be identified, indicating that the concept of self-rated health may be influenced by a large number of SNPs with very small effect size. In order to identify these, a very large sample would be needed which only can be accomplished by conducting a meta-analysis combining different samples.

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Figure 1.

The Q-Q plot shows the association between the observed and expected $-\log 10$ (P-value) of the autosomal association between SNPs and self-rated health with the grey area representing the 95% confidence interval.

1 3 5 7 6 11 10 15 10 92 4 9 10 10 15 10 921

Figure 2.

Manhattan plot showing the results of the genome-wide association analyses for self-rated health with the x-axis showing chromosome numbers and the y-axis the p-value (-log10) of the association signals.

Table 1a

Raw number and percentage (in brackets) of participants drawn from each study. The most recent score of participants who took part in more than one study was used.

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	Study 1 ^a	Study 2 ^b	Study 3 ^c	Study 4 ^d	Total
N (%)	43 (1%)	237 (4%)	158 (2%)	6268 (93%)	6706 (100%)
Age range	30–72	50-92	24–36	18-85	18–92
^a (Heath, et al.,	1997);				

 $b_{(Bucholz, et al., 1998);}$

c(Mosing, Gordon, et al., 2009);

d (Hansell, et al., 2008).

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Table 1b

Score distribution of the final sample for self-rated health.

	Excellent	Good	Fair	Poor	Total
N (%)	1837 (27%)	3640 (54%)	1053 (16%)	176 (3%)	6706 (100%)

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	ers within top 50 SNPs).
Table 2	independent marke
	n self-rated health (
	est association with
	s showing stronge
	Genetic marker

	Chr	SNP	Base pair location	p-value	SNPs in LD	Minor allele	MAF	Effect size	Closest gene	location
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2 15767380 127982645 1.5×10.6 2 A 30 -0.03 CYP27C12 156759460 127897011 3.3×10.6 4 C 24 0.03 $2YT16$ 14 156573416 62518348 3.3×10.6 1 G 24 0.03 $SYT16$ 2 152357266 924387 3.7×10.6 1 G 3.3 0.03 $SYT16$ 2 15300489 78485994 4.8×10.6 1 G 3.3 0.03 $SYT16$ 2 15300489 78485994 4.8×10.6 1 G 3.3 0.03 $SYT16$ 2 15300489 78485994 4.8×10.6 1 G 3.3 0.03 $SYT16$ 2 15300489 78485994 4.8×10.6 1 G 3.3 0.03 $SYT16$ 2 153779441 24198815 8.3×10.6 1 G 2.3 0.03 $SYT16$ 2 15179310 128176400 9.2×10.6 1 T 2.0 2.0 2.2 3 1512680321 77148730 9.3×10.6 1 T 2.2 0.03 1.03 3 159548119 38531581 9.3×10.6 1 T 2.2 0.03 1.2 3 159548119 38531581 9.3×10.6 1 A 2.2 0.03 1.2 3 159548119 38531581 9.3×10.6 1 A 2.2 0.03 1.2 <td>5</td> <td>rs17043944</td> <td>22724593</td> <td>5.3 * 10-7</td> <td>7</td> <td>Т</td> <td>.02</td> <td>-0.10</td> <td></td> <td></td>	5	rs17043944	22724593	5.3 * 10-7	7	Т	.02	-0.10		
2 rs6759460 127897011 3.3 * 10-6 1 C 24 0.03 AC110926 14 rs6573416 62518348 3.3 * 10-6 1 G .13 0.03 SYT16 2 rs5573166 9924387 3.3 * 10-6 1 G .35 -0.02 NAV3 2 rs375266 9924387 3.3 * 10-6 1 G .35 -0.02 NAV3 2 rs300489 7845994 4.8 * 10-6 1 G .35 -0.02 NAV3 2 rs300489 7845994 1.8 * 10-6 1 G .35 -0.02 NAV3 2 rs1779410 16002288 9.2 * 10-6 1 G .14 0.03 NAV3 4 rs1779810 128176040 9.2 * 10-6 1 T .23 0.02 PO012 8 rs1798810 128176040 9.2 * 10-6 1 T .41 -0.02 PO02 8 <td< td=""><td>5</td><td>rs7567389</td><td>127982645</td><td>1.5 * 10-6</td><td>2</td><td>A</td><td>.30</td><td>-0.03</td><td>CYP27C1</td><td>Upstream</td></td<>	5	rs7567389	127982645	1.5 * 10-6	2	A	.30	-0.03	CYP27C1	Upstream
14 rs6573416 62518348 3.3 * 10-6 1 G .13 0.03 SYT16 2 rs2357266 9924387 3.7 * 10-6 1 G .35 -0.02 NAV3 12 rs2357266 9924387 3.7 * 10-6 1 G .35 -0.02 NAV3 12 rs30489 7848594 4.8 * 10-6 1 G .33 0.03 NAV3 21 rs779441 24198815 8.3 * 10-6 1 G .14 0.03 NAV3 4 rs17478107 16002288 9.2 * 10-6 1 G .14 0.03 PROM1 2 rs1799810 128176400 9.2 * 10-6 1 A .20 0.03 PROM1 3 rs12680321 77148730 9.8 * 10-6 1 A .20 .003 PROM1 3 rs12680321 77148730 9.8 * 10-6 1 A .20 .003 .013 3 <t< td=""><td>5</td><td>rs6759460</td><td>127897011</td><td>3.3 * 10-6</td><td>4</td><td>C</td><td>.24</td><td>0.03</td><td>AC110926</td><td>Intronic</td></t<>	5	rs6759460	127897011	3.3 * 10-6	4	C	.24	0.03	AC110926	Intronic
2 rs3357266 9924387 3.7 * 10-6 1 G 35 -0.02 12 rs300489 7845594 4.8 * 10-6 1 G .35 -0.02 21 rs300489 7845594 4.8 * 10-6 1 G .14 0.03 NAV3 21 rs7279441 24198815 8.3 * 10-6 1 G .14 0.03 NAV3 2 rs17478107 16002288 9.2 * 10-6 7 C .28 -0.02 PROM1 2 rs179810 128176040 9.2 * 10-6 7 7 -0.02 PROM1 3 rs12680321 77148730 9.8 * 10-6 7 7 -0.02 PROM1 3 rs9548119 38531581 9.3 * 10-6 1 A .20 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 14 rs9548119 38531581 9.3 * 10-6 1 A .22	14	rs6573416	62518348	3.3 * 10-6	1	IJ	.13	0.03	SYT16	Intronic
12rs3004807848594 $4.8 * 10-6$ A230.03NAV321rs727944124198815 $8.3 * 10-6$ 1G.140.03NAV34rs1747810716002288 $9.2 * 10-6$ C.28 -0.02 PROM12rs179810128176040 $9.2 * 10-6$ T.2 -0.02 PROM18rs1268032177148730 $9.2 * 10-6$ T.2 -0.02 PROM18rs1268032177148730 $9.3 * 10-6$ 1A.20 -0.03 13rs954811938531581 $9.3 * 10-6$ 1A.20 -0.03 13rs954811938531581 $9.3 * 10-6$ 1A.20 0.03 13rs954811938531581 $9.3 * 10-6$ 1A.20 0.03 14rs954811938531581 $9.3 * 10-6$ 1A.20 0.03 15rs954811938531581 $0.3 * 10-6$ 1A.22 0.03 16rs11886712817777 $1.0 * 10-5$ C 4.1 -0.02 $ADRB216rs1181504116224241.0 * 10-55G4.90.02ADRB216rs11085795119885151.0 * 10-51A.260.03ZN43917rs11027995720754.9 * 10-51C.21-0.03ZN439$	5	rs2357266	9924387	3.7 * 10-6	1	IJ	.35	-0.02		Intergenic
21 rs779441 24198815 8.3 * 10-6 1 G .14 0.03 4 rs17478107 16002288 9.2 * 10-6 C .28 -0.02 PROM1 2 rs17478107 16002288 9.2 * 10-6 T .41 -0.02 PROM1 2 rs1799810 128176040 9.2 * 10-6 T .41 -0.02 PROM1 8 rs12680321 77148730 9.8 * 10-6 T .41 -0.02 PROM1 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 13 rs9548119 38531581 0.3 * 10-6 1 A .22 0.03 14 rs91487 1.0 * 10-5 1 A .22 0.03 16 rs	12	rs300489	78485994	4.8 * 10-6		А	.23	0.03	NAV3	Intronic
4 rs17478107 16002288 9.2 * 10-6 C 28 -0.02 PROM1 2 rs1799810 128176040 9.2 * 10-6 T 41 -0.02 PROM1 8 rs12680321 77148730 9.8 * 10-6 T 20 -0.03 13 rs9548119 38531581 9.3 * 10-6 1 A 22 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A 22 0.03 2 rs9548119 38531581 9.3 * 10-6 1 A 22 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A 22 0.03 2 rs118867 12817777 1.0* 10-5 C 41 -0.02 ADARB2 10 rs11815041 1622424 1.0 * 10-5 5 G 49 0.02 ADARB2 10 rs11085795 11988515 1.0 * 10-5 1 A 26 0.03 ZNF439	21	rs7279441	24198815	8.3 * 10-6	1	IJ	.14	0.03		Intergenic
2 rs1799810 128176040 9.2 * 10-6 T 41 -0.02 8 rs12680321 77148730 9.8 * 10-6 T 20 -0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 2 rs1158867 128177377 1.0 * 10-5 7 C .41 -0.02 10 rs11815041 1622424 1.0 * 10-5 5 G .49 0.02 ADARB2 10 rs11085795 11988515 1.0 * 10-5 1 A .26 0.03 ZNF439 10 rs11085795 1988515 1.0 * 10-5 1 A .26 0.03 ZNF439 11 rs11085795 1988515 1.0 * 10-5 1 A .26 0.03 ZNF439	4	rs17478107	16002288	9.2 * 10-6		С	.28	-0.02	PROM1	Intronic
8 rs12680321 77148730 9.8 * 10-6 T 20 -0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 2 rs1158867 12817737 1.0 * 10-5 7 41 -0.02 10 rs11815041 1622424 1.0 * 10-5 5 G .49 0.02 ADARB2 19 rs11085795 11988515 1.0 * 10-5 1 A .26 0.03 ZNF439 11 rs1102279 95720275 4.9 * 10-5 1 C .21 -0.03 MAML2	5	rs1799810	128176040	9.2 * 10-6		Т	.41	-0.02		
13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 2 rs1158867 128177377 1.0* 10-5 C .41 -0.02 10 rs11815041 1622424 1.0 * 10-5 5 G .49 0.02 ADARB2 19 rs11085795 11988515 1.0 * 10-5 1 A .26 0.03 ZNF439 11 rs7120279 95720275 4.9 * 10-5 1 C .21 -0.03 ZNF439	×	rs12680321	77148730	9.8 * 10-6		Т	.20	-0.03		Intergenic
13 rs9548119 38531581 9.3 * 10-6 1 A .22 0.03 2 rs1158867 12817737 1.0* 10-5 C .41 -0.02 10 rs11815041 1622424 1.0* 10-5 5 G .49 0.02 ADARB2 19 rs11085795 11988515 1.0 * 10-5 1 A .26 0.03 ZNF439 11 rs7120279 95720275 4.9 * 10-5 1 C .21 -0.03 MAML2	13	rs9548119	38531581	9.3 * 10-6	1	A	.22	0.03		Intergenic
2 rs1158867 12817737 1.0*10-5 C A1 -0.02 10 rs11815041 1622424 1.0*10-5 5 G A9 0.02 ADARB2 19 rs11085795 11988515 1.0*10-5 1 A 26 0.03 ZNF439 11 rs7120279 95720275 4.9*10-5 1 C .21 -0.03 MAML2	13	rs9548119	38531581	9.3 * 10-6	1	А	.22	0.03		Intergenic
10 rs11815041 1622424 1.0 * 10-5 5 G .49 0.02 ADARB2 19 rs11085795 11988515 1.0 * 10-5 1 A .26 0.03 ZNF439 11 rs7120279 95720275 4.9 * 10-5 1 C .21 -0.03 MAML2	5	rs1158867	128177377	$1.0^{*} 10^{-5}$		C	.41	-0.02		
19 rs11085795 11988515 1.0*10-5 1 A .26 0.03 ZNF439 11 rs7120279 95720275 4.9*10-5 1 C .21 -0.03 MAML2	10	rs11815041	1622424	1.0 * 10-5	5	IJ	.49	0.02	ADARB2	Intronic
11 rs7120279 95720275 4.9 * 10-5 1 C 21 -0.03 MAML2	19	rs11085795	11988515	1.0 * 10-5	1	А	.26	0.03	ZNF439	Intergenic
	Ξ	rs7120279	95720275	4.9 * 10-5	1	С	.21	-0.03	MAML2	Intronic

Twin Res Hum Genet. Author manuscript; available in PMC 2011 August 1.

= Minor Allele Frequency; Closest gene = name of gene if the SNP is located in a known gene or within 50kb distance from a gene; Base pair locations: obtained from the HapMapl+II (b36r22) CEU legend files; Closest gene to the SNP: obtained from WGA Viewer release 57. Note. Independent markers: more than 500kb apart and in LD of r² < 0.70; SNPs in LD: the number of correlated SNPs that are in the top 50 (nonindependent groups of markers); Chr = Chromosome; MAF

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Gene	Chromosome	Start position	End position	Number of SNPs in gene	P-value
ERCC3	2	127731335	127768222	74	$9.0 * 10^{-5}$
S100A5	1	151776246	151780865	30	$1.2 * 10^{-4}$
PROC	2	127892486	127903288	75	$1.5 * 10^{-4}$
S100A6	1	151773699	151775341	36	$1.7 * 10^{-4}$
S100A4	1	151782718	151784906	26	$1.8 * 10^{-4}$