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Authors
Chen, Chunhui
Chen, Wen
Chen, Chuansheng
et al.

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Peer reviewed
Genetic Variations in the Serotonergic System Contribute to Body-Mass Index in Chinese Adolescents

Chunhui Chen1*, Wen Chen1*, Chuansheng Chen2, Robert Moyzis3, Qinghua He4, Xuemei Lei1, Jin Li1, Yunxin Wang1, Bin Liu1, Daiming Xiu1, Bi Zhu1, Qi Dong1*

1 State Key Laboratory of Cognitive Neuroscience and Learning, Beijing Normal University, Beijing, China, 2 Department of Psychology and Social Behavior, University of California Irvine, Irvine, California, United States of America, 3 Department of Biological Chemistry and Institute of Genomics and Bioinformatics, University of California Irvine, Irvine, California, United States of America, 4 Department of Psychology, University of Southern California, Los Angeles, California, United States of America

Abstract

Objective: Obesity has become a worldwide health problem in the past decades. Human and animal studies have implicated serotonin in appetite regulation, and behavior genetic studies have shown that body mass index (BMI) has a strong genetic component. However, the roles of genes related to the serotonergic (5-hydroxytryptamine, 5-HT) system in obesity/BMI are not well understood, especially in Chinese subjects.

Subjects and Design: With a sample of 478 healthy Chinese volunteers, this study investigated the relation between BMI and genetic variations of the serotonergic system as characterized by 136 representative polymorphisms. We used a system-level approach to identify SNPs associated with BMI, then estimated their overall contribution to BMI by multiple regression and verified it by permutation.

Results: We identified 12 SNPs that made statistically significant contributions to BMI. After controlling for gender and age, four of these SNPs accounted for 7.7% additional variance of BMI. Permutation analysis showed that the probability of obtaining these findings by chance was low (p = 0.015, permuted for 1000 times).

Conclusion: These results showed that genetic variations in the serotoninergic system made a moderate contribution to individual differences in BMI among a healthy Chinese sample, suggesting that a similar approach can be used to study obesity.


Introduction

A decade ago, the World Health Organization warned about a growing obesity epidemic and listed more than 30 diseases that are causally related to obesity [1]. Globally, approximately 1.6 billion adults are either overweight (BMI [weight in kilogram divided by the square of height in meter] ≥ 25) or obese (BMI ≥ 30) [2]. In fact, the rates of obesity have tripled in developing countries in the past 20 years [3]. Moreover, childhood obesity is also increasing rapidly worldwide [4].

Although many environmental factors (e.g., freely available high-calorie food, sedentary life style, low socio-economic status and high-danger neighborhood environment) predispose individuals to gaining weight [5,6,7,8], genetic factors also contribute to energy homeostasis or appetite, which can lead to obesity. Family, twin, and adoption studies indicate that 24%–90% of human BMI variation is due to genetic factors [9,10,11,12,13,14]. Recent molecular genetic studies have identified many genes that regulate appetite or energy balance (e.g., FTO, MC4R, SH2B1, and serotonin related genes) and have robust associations with obesity or BMI [15,16].

Because serotonin can regulate appetite by activating pro-opiomelanocortin (POMC) neurons, which play a key role in the regulation of feeding by sending anorectic signals to the periventricular nucleus (PVN) and other brain areas associated with energy homeostasis [17], serotonin as well as related genes are often tested for association with weight gain and obesity. Indeed, a strong negative correlation between blood 5-HT concentration and body mass was found both in mice [18] and in human [19]. Studies of SERT knockout mice have uncovered SERT as a candidate gene for obesity, with SERT mutant (SCL6A4/−/−) mice becoming obese [20]. This polymorphism has also been associated in some studies with eating disorder [21,22,23] and obesity [24,25], although other studies showed no association between the 5-HTTLPR polymorphism and weight regulation [26,27,28]. In terms of the 5-HT receptor genes, the serotonin (5-HT) receptor HTR2C was demonstrated to play a role
in modulating appetite behavior using knock-out mice [29,30],
normal population [31] and patients [32,33,34,35], although some
studies [36,37] failed to replicate that result. HTR1B [38,39],
HTR2A [40,41], HTR3B [42] were also reported to be associated
with body mass or obesity. MAO-A was also found to influence body
mass [43] or obesity [44].

Although these serotonin-related genes have been identified as
being relevant to body mass and obesity, the results have not
always been consistent and the size of their effects has been
typically small, far less than previously estimated 24–90%
heritability. There may be many reasons for these inconsistent
results and small effect sizes. One most likely reason is
polygenicity. Complex quantitative traits are influenced by many
genes, each with a small effect. As early as 1918, Fisher proposed
this polygenic model that combined many genes of small effects to
yield the continuous variation for most quantitative traits [45].
Recently, some studies have successfully applied the polygenic
model by combining effects of the whole genome [46,47,48] or
effects of genes within a pathway [49,50,51]. Since several
serotonin-related genes exert their effect on BMI, it is likely that
their effects are cumulative. The current study used a system-level
approach to examine the role of the serotoninergic system in
BMI/obesity.

Another possible reason for inconsistent results may be the
heterogeneity in samples across studies. Subjects in different
studies differ in their health status, age, sex, and ethnicity, which
might have confounded the relations between genes and BMI. For
example, associations between 5HT2A and BMI are found in
obese [40] and anorexia nervosa patients [41] but not in healthy
controls. Similarly, 5HTTLPR was associated with BMI in non-
elderly (<65 yr) stroke patients but not in elderly patients (> or = 65 yr).
An association was observed between MAO-A and obesity among white and Hispanic American subjects, but not among African-American subjects [44]. Thus it is important to
control for these potential confounding factors.

The current study adopted the system-level approach to
examine the role of the serotoninergic system in body mass in
a relatively homogenous sample (in terms of age, health status, and
ethnicity). We enrolled a sample of young healthy Han Chinese
subjects, genotyped polymorphisms within the serotonin system,
and calculated their BMI. Specifically, we selected 136 poly-
morphic loci (including 134 SNPs and 2 VNTR polymorphisms)
to cover a substantial portion (by LD) of the common variations
within known genes of the 5-HT system to estimate the additive
and multiplicative contributions of these genes on BMI.

Materials and Methods

Participants

Four hundred and eighty healthy Chinese college students
(mean age = 20, SD = 1) were recruited from Beijing Normal
University, Beijing, China. They had normal or corrected-to-
normal vision, and had no history of neurological or psychiatric
problems according to self-report. None of them were identified as
having alcohol or nicotine dependence according to the Alcohol
Use Disorders Identification Test [52] and the Fagerström Test for
Nicotine Dependence [53]. Two participants were excluded
because of poor genotyping results. A written consent form was
obtained from each subject after a full explanation of the study
procedure. This study was approved by the IRB of the State Key
Laboratory of Cognitive Neuroscience and Learning at Beijing
Normal University, China.

BMI Measurements

Height and weight of subjects were self-reported. BMI was
calculated as weight (kg) divided by the square of height (m). Self-
reported data on weight and height have been used by previous
large-scale studies on body mass and proved to be highly reliable
in calculating BMI [22,38,46,54,55,56,57]. Furthermore, all
students including all of our participants were given an annual
physical examination at the beginning of the academic year in
September and they were informed of their height and weight.
Self-report data on height and weight were collected in December.

Genetic Analysis

Gene selection. We selected 25 genes and 136 associated
polymorphisms (134 SNPs and 2 VNTR polymorphisms) distrib-
uted across the synthesis, degradation, transporter, and receptor
subsystems of the 5-HT system. 5-HT synthesis involves convert-
ing the tryptophan (via TPH) to 5-hydroxytryptophan (5-HTP),
followed by subsequent hydroxylation (by TPH) to 5-HT. We
included two genes related to 5-HT synthesis: tryptophan
hydroxylase (TPH1 and TPH2, with three SNPs each). For the
degradation subsystem, released 5-HT is directly broken down at
the synapse into inactive metabolites by two enzymes, COMT and
MAO (including MAO-A and MAO-B). We included catechol-O-
methyl transferase gene (COMT, with 7 SNPs) and monoamine
oxidase genes (MAOA, with 3 SNPs and 1 VNTR, and MAOB
with 3 SNPs). The 5-HT transporter includes (1) SLC6A4, an integral
membrane-spanning protein that pumps the neurotransmitter
serotonin from synaptic spaces into presynaptic neurons and (2)
VMAT, a transport protein integrated into the membrane of
intracellular vesicles of presynaptic neurons, which acts to
transport monoamines into the synaptic vesicles. We included
SLC6A4 (7 SNPs plus 5HTTLPR), VMAT1 (SLC18A1, 9 SNPs),
and VMAT2 (SLC18A2, 5 SNPs). For the receptor subsystem,
we included all 17 genes (with the respective number of SNPs in
parentheses); HTR1A (2), HTR1B (2), HTR1D (13), HTR1F (5),
HTR2A (21), HTR2B (6), HTR2C (3), HTR3A (1), HTR3B (2),
HTR3C (3), HTR3D (4), HTR3E (2), HTR4 (10), HTR5A (4),
HTR5B (2), HTR6 (5), and HTR7 (7). Together, the above 25
genes represent all major genes involved in these four 5-HT
subsystems in humans [58]. Details about these genes and the
selected loci can be found in Table S1.

Genotyping techniques. The SNPs were genotyped using the
standard Illumina Golden Gate Genotyping protocol (see Illumina
southgene.com.cn; Shanghai South Gene Technology Co., Ltd,
Shanghai, China). In addition, three genetic markers (5HTTLPR,
MAOA VNTR, and COMT rs4680) were ascertained by standard
PCR procedures [59,60,61].

Gene data preprocessing. Two subjects with more than
10% null genotyping were excluded. In addition to automatic
calling of genotypes, Illumina genotyping platform supplied a
quantitative quality measure known as the GenCall score. It
measures how close a genotype is to the center of the cluster of
other samples assigned to the same genotype, compared with
the centers of the clusters of the other genotypes. This measure
ranges from 0 to 1, with a higher score indicating a more reliable result.
The conventional cutoff point is.25 [62]. Of the 63574 genotypes
(133 SNPs by 478 subjects) in the current study, 120 genotypes
(0.2%) were excluded because their GenCall scores were lower
than .25.

Additional data cleaning included the treatment of low-
frequency alleles. For SNPs with either heterozygote or minor
homozygote found in fewer than 10 (about 2%) participants, these
two genotype groups were combined. If the combined group still
had fewer than 10 participants, the SNPs were excluded from further analysis. SNPs that showed no polymorphisms were also deleted. In order to examine sample representativeness, Hardy-Weinberg equilibrium (HWE) index was calculated using the Chi square test and setting $\alpha$ to 1. Since males have only one X chromosome, only females were included in HWE calculation for SNPs located on X chromosome. Five of the SNPs showed significant HW disequilibrium ($p<0.01$). The inclusion of both SNPs and additional SNPs in regions detected in selection screens [63,64] resulted in high LD among a number of SNPs. Thirty-one SNPs included in initial analysis were excluded from multiple regression analysis because of their high LD with other adjacent SNPs ($R^2>0.5$, calculated with Plink [63]), yielding a final list of 105 polymorphisms for the main data analyses. It is worth mentioning that the “redundant” SNPs showed the same or almost the same results as the linked SNPs, confirming the association. Table S1 shows the details about all 136 polymorphic loci (134 SNPs and 2 VNTRs) included in our study: location (rs number, chromosome, position), gene, serotonin subsystem, allele polymorphism and frequency, Hardy Weinberg equilibrium, linkage disequilibrium and deleted SNPs. Finally, genetic relatedness of subjects was checked following Anderson et al. [66] protocol using Plink. We used all 240 unrelated autosome SNPs ($r^2<0.8$) available in the larger project of these subjects and set the threshold of 0.95 (personal communication with Drs. Anderson and Zondervan). We found no pair of subjects showing high relatedness (all PI_HAT smaller than or equal to 0.5).

Data Analysis

The goal of the current study was to understand the relation between individual differences in BMI and genetic variations in the 5-HT system in healthy subjects. Moving beyond the single-gene or a small number of haplotypes approaches used in typical molecular behavior genetics research, this study used the system-level approach [50] to examine the overall contributions of the serotonergic system (characterized by the major genes and their associated loci) on BMI.

Briefly, the analysis includes three steps: First, ANOVA was used to screen polymorphism loci that showed nominal significance ($p<0.05$) on BMI; these loci were then entered into a regression model to estimate their overall contribution to BMI after controlling for gender and age; and lastly the regression model was verified by permutation. In this study, we built two kinds of regression models. In model 1 (main effects), we included the loci with significant main effects based on the ANOVA results and used the forward stepwise method to build the model. Gender and age were entered as control variables. To run multiple regression analyses, all SNPs were coded in a linear way, i.e. the major homozygote, heterozygote, and minor homozygote were coded as 1, 2, and 3, respectively (SNPs on X chromosome were coded as 1 and 3 for major and minor allele, and 2 for female heterozygotes). In addition, the MAOA VNTR was coded as 1 for the 3 repeat and 3 for the 4 repeat in males and 1 for 3 repeat homozygotes and 3 for others in females. In model 2, all two-way interactions of these loci in model 1 were added using forward stepwise method. Permutation was done 1000 times by shuffling BMI (along with gender and age) across subjects, and the probability of getting a larger $R^2$ in the shuffled data than in the real data was defined as $p$ value of the model.

Results

The mean BMI for our sample was 20.5 kg/m$^2$ (SD = 2.4), ranging from 16.3 to 37.5. According to WHO BMI classification, there were 93 (71 female) underweight participants (BMI <18.5), 359 (192 female) normal weight participants (18.5 ≤ BMI <25), and 26 (8 female) overweight participants (BMI ≥25). The BMI distribution in the present study was comparable to other studies with Chinese college students [67,68]. Males (21.14±2.44) had significantly higher BMI than females (20.00±2.67; $t(476) = 5.30$, $p = 1.00 \times 10^{-7}$), which was consistent with previous findings in healthy young Chinese [69].

Of the 105 SNPs, 12 showed significant main effects with uncorrected $p<0.05$. Specifically, individuals with the following genotypes showed lower BMI than those with alternative alleles: homozygous for the major allele of rs13166761 (HTR4), rs11214769 (SLC18A1 (VMAT1)), rs11214769 (HTR3B), rs977003 (HTR2A), rs2224721 (HTR2A), rs2192371 (HTR2C), rs4911871 (HTR2C), or rs2270638 (VMAT1); or heterozygous/ minor allele homozygous for rs6651806 (MAOB), rs5905512 (MAOB); or homozygous for the major allele of rs2904569 (HTR7) or rs6644065 (HTR2C) (see Table 1, and Table S2 for effects of all loci). These SNPs were used in a regression analysis to build model 1 (main effects). There was no significant gender-by-SNP interaction except rs5905512 (see Table 1), and this SNP did not contribute to regression model 1, so we included gender, but not gender-by-gene interactions, as a covariate in the following analysis.

Table 2 shows the results of the multiple regression analysis. On the first step, two control variables (gender and age) were entered. Together they accounted for 5.6% variance of BMI. On the second step, forward stepwise regression resulted in four of the 12 SNPs to be included in the regression equation, showing that they made unique contributions to explaining variance in BMI. Together these SNPs accounted for 7.7% additional variance, yielding a total $R^2$ of 13, $F(6,455) = 11.61, p = 4.08 \times 10^{-12}$. Permutation results are shown in Figure 1. Based on 1000 permutations, the probability of attaining the $R^2$ or adjusted $R^2$ found in our model was 0.015 and 0.011, respectively. We then added potential interactive effects to investigate whether additional variance in BMI can be accounted for by gene–gene interactions. In this analysis, we first entered the control variables (gender and age) and the four SNPs in model 1 and finally their two-way interactions using the stepwise procedure. For the four SNPs that entered model 1, there were 6 potential interactions. None of the interaction terms made significant and unique contributions to the model.

Discussion

Based on the system-level analysis of 5-HT neurotransmitter genes, we identified 12 SNPs of the 5-HT-related genes showing nominal effects on BMI. Four of these SNPs made significant unique contributions to BMI even after controlling for gender and age. This result has two significant implications. First, the current study revealed a significant role for genes in the 5-HT system on BMI among Chinese, confirming that body mass is likely to be understood under the genetic basis of a complex trait such as body mass. This approach can estimate the overall contribution of genes within a pathway and can help to explain the missing heritability [46].

We found that 12 SNPs of seven genes (MAOB, SLC18A1 (VMAT1), HTR2A, HTR2C, HTR3B, HTR4, and HTR7) were significantly associated with BMI. As summarized in the introduction, previous studies have already found evidence, although
not always consistent, of association between the HTR2A and HTR2C genes and BMI. However, other genes we identified have not been tested previously in BMI-related studies to the best of our knowledge.

VMAT1 is expressed primarily in neuroendocrine cells such as the adrenal medulla and pineal gland [70,71,72]. As early as in 1999, Hayashi et al. [73] found that VMAT1 was responsible for the storage of 5-hydroxytryptamine in rat pinealocytes. Mammalian pinealocytes contain more 5-HT than any other cells. Upon stimulation by norepinephrine (NE), the internal 5-HT is released and then stimulates serotonin N-acetyltransferase activity via the 5-HT2 receptor, resulting in increased melatonin output [73].

Serotonin (5-HT) has been found to be involved in energy metabolism and then stimulates serotonin N-acetyltransferase activity via the 5-HT2 receptor, resulting in increased melatonin output [73]. The VMAT1 gene was found to be associated with anxiety-related personality traits [77] and anxiety has been shown to be associated with obesity [78,79] or BMI [80]. Previous studies have found that VMAT plays an important role in the life cycle of ghrelin and obestatin in the A-like cells of the stomach [81,82], and ghrelin and obestatin have effects on food intake and energy balance. Therefore, we speculate that the VMAT1 gene may have an effect on BMI through melatonin output, ghrelin, obestatin or anxiety mood. This gene accounted for the largest proportion of the variance of BMI in our study (Table 2).

The 5-HT3 receptor has been suggested to be involved in anxiety, depression, pain, alcohol dependence, and eating disorders [42,83]. The HTR2C gene encodes the B-subunit of the type 3 serotonin receptor (5-HT3), a ligand-gated ion channel that is known to be involved in gut motility and peristalsis. Thus the HTR3B gene may regulate BMI because gut motility is associated with numerous gastrointestinal peptides with significant effects on food intake and energy balance [84]. Many studies have also reported that the 5-HT3B polymorphism is associated with the incidence of major depression [85], efficiency of the antidepressant treatment [86], and the incidence and severity of nausea after paroxetine treatment of psychiatric patients [87]. Although the specific biological mechanisms are not well understood, our results indicate that HTR3B gene polymorphism may influence gut motility or mood.

Our analysis also showed that HTR2C and HTR2A are possible factors influencing BMI in Chinese subjects, as have been reported by previous studies. Different from the most often studied C759T polymorphism associated with weight gain [31,32,33,34,35,88], three SNP we found related to BMI are all located in the intron region of HTR2C. First, there is strong evidence for an interaction between leptin and the 5-HTergic system [88]. Second, McCarthy et al. showed a strong effect of HTR2C polymorphism −759G>A on circulating leptin levels after adjusting for body fat. Other studies also suggested that serotonin influences food intake because of variations in the HTR2C receptor [89,90]. Similarly, previous researchers have also found an association between a polymor-

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### Table 1. Means and standard deviations of BMI for each polymorphism, and main effects and post hoc comparisons of SNPs that showed significant main effects and were used in subsequent multiple regression analysis.

<table>
<thead>
<tr>
<th>SNP</th>
<th>Subsystem</th>
<th>Gene</th>
<th>Maj Mean±SD</th>
<th>n</th>
<th>Het Mean±SD</th>
<th>n</th>
<th>Min Mean±SD</th>
<th>n</th>
<th>F</th>
<th>p</th>
<th>mh</th>
<th>mm</th>
<th>hm*</th>
<th>F(p)</th>
</tr>
</thead>
<tbody>
<tr>
<td>rs6651806</td>
<td>Degradation</td>
<td>MAOB</td>
<td>20.63 2.45</td>
<td>38</td>
<td>19.95 2.16</td>
<td>97</td>
<td>CC</td>
<td>6.32</td>
<td>0.01</td>
<td>0.49 (0.53)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>rs5905512</td>
<td></td>
<td>MAOB</td>
<td>20.76 2.62</td>
<td>284</td>
<td>20.10 2.01</td>
<td>194</td>
<td>GG</td>
<td>9.02</td>
<td>0.01</td>
<td>0.02 (0.89)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>rs1018079</td>
<td>Transport</td>
<td>SLC18A1</td>
<td>20.29 2.32</td>
<td>303</td>
<td>20.72 4.35</td>
<td>18</td>
<td>CC</td>
<td>5.92</td>
<td>0.01</td>
<td>0.02 (0.71)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>rs2270638</td>
<td></td>
<td>SLC18A1</td>
<td>20.33 2.47</td>
<td>344</td>
<td>20.92 2.33</td>
<td>133</td>
<td>CC</td>
<td>5.77</td>
<td>0.02</td>
<td>0.27 (0.60)</td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>rs977003</td>
<td>Receptor</td>
<td>HTR2A</td>
<td>20.29 2.30</td>
<td>299</td>
<td>20.92 2.19</td>
<td>151</td>
<td>CC</td>
<td>3.45</td>
<td>0.01</td>
<td>0.14 (0.87)</td>
<td></td>
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<td></td>
<td></td>
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<tr>
<td>rs2224721</td>
<td></td>
<td>HTR2A</td>
<td>20.12 2.12</td>
<td>216</td>
<td>20.79 2.51</td>
<td>53</td>
<td>AA</td>
<td>4.85</td>
<td>0.01</td>
<td>0.82 (0.48)</td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>rs2192371</td>
<td></td>
<td>HTR2C</td>
<td>20.61 2.42</td>
<td>242</td>
<td>19.79 1.80</td>
<td>124</td>
<td>CC</td>
<td>8.62</td>
<td>0.01</td>
<td>0.01 (0.95)</td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>rs6644085</td>
<td></td>
<td>HTR2C</td>
<td>20.45 2.26</td>
<td>373</td>
<td>20.05 2.81</td>
<td>68</td>
<td>GG</td>
<td>7.14</td>
<td>0.01</td>
<td>0.16 (0.01)</td>
<td></td>
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<tr>
<td>rs4911871</td>
<td></td>
<td>HTR2C</td>
<td>20.39 2.23</td>
<td>350</td>
<td>20.17 2.74</td>
<td>76</td>
<td>AA</td>
<td>7.44</td>
<td>0.01</td>
<td>0.12 (0.30)</td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>rs11214769</td>
<td></td>
<td>HTR3B</td>
<td>20.31 2.22</td>
<td>335</td>
<td>20.88 2.85</td>
<td>124</td>
<td>GG</td>
<td>3.46</td>
<td>0.03</td>
<td>0.58 (0.19)</td>
<td></td>
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<td></td>
<td></td>
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<tr>
<td>rs13366761</td>
<td></td>
<td>HTR4</td>
<td>20.38 2.23</td>
<td>244</td>
<td>20.79 2.64</td>
<td>202</td>
<td>AA</td>
<td>4.74</td>
<td>0.01</td>
<td>0.02 (0.59)</td>
<td></td>
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<td></td>
<td></td>
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<tr>
<td>rs7904569</td>
<td></td>
<td>HTR7</td>
<td>20.50 2.38</td>
<td>206</td>
<td>20.69 1.83</td>
<td>215</td>
<td>CC</td>
<td>3.64</td>
<td>0.03</td>
<td>0.13 (0.26)</td>
<td></td>
<td></td>
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<td></td>
</tr>
</tbody>
</table>

Note: Empty cells mean no such genotypes were found in our sample. Maj: Major allele; Het: Heterozygote; Min: Minor allele.

*Results (p values) of post hoc comparisons. mh = Maj versus Het, mm = Maj versus Min, hm = Het versus Min.

**Post hoc comparison was not run because there were only 2 groups for this locus.

* doi:10.1371/journal.pone.0058717.t001

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### Table 2. Regression models.

<table>
<thead>
<tr>
<th>Regressor</th>
<th>Gene</th>
<th>Beta</th>
<th>T</th>
<th>P</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gender</td>
<td></td>
<td>0.24</td>
<td>−5.39</td>
<td>0.00</td>
</tr>
<tr>
<td>Age</td>
<td></td>
<td>0.02</td>
<td>0.47</td>
<td>0.64</td>
</tr>
<tr>
<td>rs1018079</td>
<td>SLC18A1/VMAT1</td>
<td>0.16</td>
<td>3.63</td>
<td>0.00</td>
</tr>
<tr>
<td>rs11214769</td>
<td>HTR3B</td>
<td>0.14</td>
<td>3.08</td>
<td>0.00</td>
</tr>
<tr>
<td>rs2224721</td>
<td>HTR2A</td>
<td>0.12</td>
<td>2.76</td>
<td>0.01</td>
</tr>
<tr>
<td>rs4911871</td>
<td>HTR2C</td>
<td>0.12</td>
<td>2.81</td>
<td>0.01</td>
</tr>
</tbody>
</table>

Note: ‘Gene’ is the corresponding gene for each SNP; ‘beta’ is the standardized regression coefficient, ‘T’ and ‘P’ are t-test results.

* doi:10.1371/journal.pone.0058717.t002
In conclusion, we used a system-level approach to identify several genetic SNPs associated with variations in BMI. This analysis provides further evidence for the association between genetic variants in the serotonin pathway and BMI. Because current lifestyle interventions are largely ineffective in addressing the challenges of growing obesity [90,99], new insights into the biology of obesity are critically needed to guide the development and application of future therapies and interventions.

Supporting Information

Table S1 Detailed information of the loci used in this study.

Table S2 Means and standard deviations of BMI for each polymorphism, and main effects and post hoc comparisons of each locus.

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Author Contributions

Conceived and designed the experiments: Chunsheng Chen RM Chunhui Chen QD. Performed the experiments: QH XL JL YW BL DX RZ. Analyzed the data: Chunhui Chen WC. Wrote the paper: WC Chunhui Chen Chunsheng Chen.

References


Figure 1. Permutation results for $R^2$ (left panel) and adjusted $R^2$ (right panel). Dashed line represents distribution of $R^2$ obtained from randomized data and solid line represents the observed $R^2$. doi:10.1371/journal.pone.0058717.g001


