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EMR-linked GWAS study: investigation of variation landscape of loci for body mass index in children

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Common variations at the loci harboring the fat mass and obesity gene (FTO), MC4R, and TMEM18 are consistently reported as being associated with obesity and body mass index (BMI) especially in adult population. In order to confirm this effect in pediatric population five European ancestry cohorts from pediatric eMERGE-II network (CCHMC-BCH) were evaluated.

Method: Data on 5049 samples of European ancestry were obtained from the Electronic Medical Records (EMRs) of two large academic centers in five different genotyped cohorts. For all available samples, gender, age, height, and weight were collected and BMI was calculated. To account for age and sex differences in BMI, BMI z-scores were generated using 2000 Centers of Disease Control and Prevention (CDC) growth charts. A Genome-wide association study (GWAS) was performed with BMI z-score. After removing missing data and outliers based on principal components (PC) analyses, 2860 samples were used for the GWAS study. The association between each single nucleotide polymorphism (SNP) and BMI was tested using linear regression adjusting for age, gender, and PC by cohort.

Results: The mean age of subjects was 9.8 years (range 2–19). The proportion of male subjects was 56%. In these cohorts, 14% of samples had a BMI ≥95 and 28 ≥85%. Meta analyses produced a signal at 16q12 genomic region with the best result of $p = 1.43\times10^{-7}$ ($D_{\text{meta}} = 7.34 \times 10^{-5}$) for the SNP rs8050136 at the first intron of FTO gene ($z = 5.26$) and with no heterogeneity between cohorts ($p = 0.77$). Under a recessive model, another published SNP at this locus, rs1421085, generates the best result ($z = 5.792$, $D_{\text{meta}} = 8.21 \times 10^{-5}$). Imputation in this region using dense 1000-Genome and Hapmap CEU samples revealed 71 SNPs with $p < 10^{-5}$, all at the first intron of FTO locus. When heterogeneity was permitted between cohorts, signals were also obtained in other previously identified loci, including MC4R (rs12964056, $p = 6.87 \times 10^{-7}$, $z = -4.98$), cholecystokinin CCK (rs8192472, $p = 1.33 \times 10^{-6}$, $z = -4.85$), Interleukin 15 (rs2099884, $p = 1.27 \times 10^{-5}$, $z = 4.34$), low density lipoprotein receptor-related protein 1B (LRP1B (rs7583748, $p = 0.00013$, $z = -3.81$)) and near transmembrane protein 18 (TMEM18) (rs7561317, $p = 0.001$, $z = -3.17$). We also detected a novel locus at chromosome 3 at COL6AS1 (best SNP = rs1542829, minor allele frequency (MAF) of 5%, $p = 4.35 \times 10^{-9}$, $z = 5.89$).

Conclusion: An EMR linked cohort study demonstrates that the BMI-Z measurements can be successfully extracted and linked to genomic data with meaningful confirmatory results. We verified the high prevalence of childhood rate of overweight and obesity in our cohort (28%). In addition, our data indicate that genetic variants in the first intron of FTO, a known adult genetic risk factor for BMI, are also robustly associated with BMI in pediatric population.

Keywords: BMI, obesity, polymorphism, GWAS
INTRODUCTION

The electronic MEdical Records and Genomics (eMERGE) Network, founded in 2007, is a consortium of multiple adult and pediatric institutions developed to explore the utility of DNA bio repositories linked to electronic medical records (EMR) in advancing genomics medicine (McCarty et al., 2011). For each site, the primary site-specific phenotypes have undergone genome-wide association studies (GWAS) with data and results shared through the network. In the pediatric population, however, genetic studies are challenging due to different developmental phase and growth patterns, different spectrums of disease, and unusual rate genetic or congenital abnormalities.

In both adults and children, obesity is a major risk factor for a number of chronic diseases, a steady and continuous rise in prevalence over the past four decades holds serious and ominous medical and economic burdens (Kopelman, 2000). The phenotype is highly heritable. Family and twin studies have shown that between 40 and 70% of the inter-individual variation in obesity can be attributable to genetic factors (Maes et al., 1997). In recent years, large-scale GWAS have identified many loci associated with Body Mass Index (BMI), the most common measure of obesity, but these loci combined explain only 2–4% of the heritability (Speliotes et al., 2010). Thus far, four waves of GWAS studies for BMI identified 32 loci that reached genome-wide significance and unequivocally were associated with BMI in a large meta-analysis performed by the GIANT (Genetic Investigation of ANtropometric Traits) consortium (Frayling et al., 2007; Scuteri et al., 2007; Loos et al., 2008; Willer et al., 2009; Speliotes et al., 2010; Loos, 2012; Magi et al., 2013). The firstly identified locus, FTO (fat mass variation in BMI (0.34%; Speliotes et al., 2010; Loos, 2012).

As a consequence, of all 32 BMI-associated loci, the frequency of the BMI-increasing allele is high in white Europeans and obesity associated gene), has the largest effect on obesity-susceptibility with obesity at a risk of 1.20 fold. Moreover, the phenotype is highly heritable. Family and twin studies have shown that between 40 and 70% of the inter-individual variation in obesity can be attributable to genetic factors (Maes et al., 1997). In recent years, large-scale GWAS have identified many loci associated with Body Mass Index (BMI), the most common measure of obesity, but these loci combined explain only 2–4% of the heritability (Speliotes et al., 2010). Thus far, four waves of GWAS studies for BMI identified 32 loci that reached genome-wide significance and unequivocally were associated with BMI in a large meta-analysis performed by the GIANT (Genetic Investigation of ANtropometric Traits) consortium (Frayling et al., 2007; Scuteri et al., 2007; Loos et al., 2008; Willer et al., 2009; Speliotes et al., 2010; Loos, 2012; Magi et al., 2013). The firstly identified locus, FTO (fat mass variation in BMI (0.34%; Speliotes et al., 2010; Loos, 2012).

High throughput single nucleotide polymorphism (SNP) genotyping was carried out previously in CCHMC and BCH using different IlluminaTM or AffymetrixTM platforms (Table 1). Quality control (QC) of the data was performed before imputation. In each genotyped cohort, standard QC criteria were met and SNPs were removed if (a) >10% missing genotyping, (b) out of Hardy–Weinberg equilibrium (HWE, P < 0.001), or a minor allele frequency (MAF) <1%. Samples with call rate <98% were excluded. Principle component analysis (PCA) was performed to identify outliers and hidden population structure using EIGENSTRAT (Price et al., 2006). Based on examination of the score plots, the first two PCs were retained and used as covariates during the association analysis in order to adjust for population stratification.

PHENOTYPING

We obtained height and weight measurements from EMR in order to calculate BMI [wt (kg) / (ht (m))2]. When multiple measurements were available for a subject, the most recent measurement was selected and all inconsistent measures were excluded. BMI z-scores and percentiles were generated using the 2000 CDC growth charts (study1). These z-scores and percentiles account for the age and sex differences in BMI throughout childhood. All data for BMI z scores (~3 to +3, mean = 0) were scaled to positive value (+4, 1–7, mean = 4) to be used as a quantitative trait for the GWAS study. In order to assess the burden of increased body weight on health and estimate the effect size, standard cut-offs were also used, and the tail BMI distribution was considered as a binary phenotype (≥95% as case and ≤20% as control). For the published FTO locus, Phenome wide association study (PheWas) was also performed in which presence or absence of each ICD-9 codes were considered as binary phenotype. Only ICD-9 codes with 50 or more available samples were included (143 codes) in the analysis.

STATISTICAL ANALYSIS

Genome-wide association studies analysis was performed by cohort in PLINK (Purcell et al., 2007) using regression models and adjusting for age, sex, and the first two principal components (PC). For BMI z-score, the primary analysis was performed using an additive effect of the minor alleles. However, as previous BMI associations have reported better model fit with different models, recessive and dominant models were subsequently evaluated. For binary phenotypes (dichotomization of BMI using a tails approach and PheWas analyses), allelic association was assessed between cases and controls by chi-square with 1 degree of freedom (df). Allelic odds ratio (OR) and 95% confidence intervals (95% CI) were obtained. In PheWas analyses, permutation procedure was performed using sample randomization strategy in which case

(http://www.cdc.gov/healthyweight/downloads/BMI_group_calculator_English.xls)
and control labels are permuted randomly (%10000) in order to obtain empirical p-values and to correct for multiple testing.

For specific target regions, imputation-based analyses were performed using the impute2-Gtool pipeline and the publicly available 1000 Genomes Project as the reference haplotype panel composed of 1092 samples (release version 2 of the 1000 Genomes Project Phase II (Howie et al., 2011). For each batch of imputation runs, the standard Markov Chain Monte Carlo (MCMC) algorithm implemented in impute-2, was used with the following threshold criteria (burnin = 10, iteration = 30, and Ne = 20000, buffer = 250 kb). A threshold of 0.90 for the posterior probability of each genotype was then applied for genotype calling and conversion using Gtool. For each imputation run, the overall genotype concordance rate was more than 95%. Additional post imputation filtering were also implemented to remove poorly imputed variants with low concordance rate according to the impute-2 standard protocol (info >0.4, Howie et al., 2011). To graphically display the results, LocusZoom was used (Pruim et al., 2010).

META-ANALYSES

The results from primary analysis in each cohort were assembled to conduct a fixed effects weighted Z meta-analysis using Meta (Willer et al., 2010). This approach controls the differences in phenotype scaling across the studies and weights the signed p-values and to correct for multiple testing. For each batch of imputation runs, the standard Markov Chain Monte Carlo (MCMC) algorithm implemented in impute-2, was used with the following threshold criteria (burnin = 10, iteration = 30, and Ne = 20000, buffer = 250 kb). A threshold of 0.90 for the posterior probability of each genotype was then applied for genotype calling and conversion using Gtool. For each imputation run, the overall genotype concordance rate was more than 95%. Additional post imputation filtering were also implemented to remove poorly imputed variants with low concordance rate according to the impute-2 standard protocol (info >0.4, Howie et al., 2011). To graphically display the results, LocusZoom was used (Pruim et al., 2010).

Table 1 | Demographic distribution of pediatric cohorts under study.

| Cohort | # Europeans | # After removing outliers and missing values | Mean age (95% CI) | Array |
|--------|-------------|---------------------------------------------|-------------------|-------|
| CHB    | 2860        | 2860                                        | 13.30 (12.97–13.66)| Omni-5|
| CCHMC  | 1603        | 1603                                        | 9.8 (9.56–10.05)  | Affymetrix-Axiom |
| Total  | 5049        | 5049                                        | 9.8 (9.67–10.65)  |       |

Table 2 | Summary of BMI-for-age and prevalence of overweight and obese children from the CCHMC-BCH cohorts.

| Male | Female | Total |
|------|--------|-------|
| 1603 | 1257   | 2860  |
| 10   | 9      | 19    |
| 62   | 62     | 124   |
| 28   | 28     | 56    |
| 15   | 12     | 27    |

Number of children assessed 2860

Underweight (<5th %ile) 10

Normal BMI (6th–40th %ile) 62

Overweight or obesity (≥40th %ile)* 28

Obese (≥95th %ile) 15

Summary of children’s BMI-for-age and prevalence of overweight and obesity

*Terminology based on Barlow and the Expert Committee (2007).
**FIGURE 1** (A,B) Manhattan plot and Q–Q plot of SNP markers used for meta-analyses (genomic inflation, $\lambda = 1.007$).

**Table 3** Most associated SNPs with BMI-$z$ in CCHMC-BCH pediatric cohorts.

| Chr | Position | Gene | SNP  | Minor allele | MAF  | $Z$ score $^*$ | $P_{\text{add}}$ | $P_{\text{rec}}$ | $P_{\text{Dom}}$ | Cochran-q | $I^2$ (%) |
|-----|----------|------|------|--------------|------|--------------|----------------|----------------|----------------|------------|----------|
| 16  | 53816275| FTO  | rs8050136 | A  | 0.39 | 5.26 | 1.43e−07 | 734e−08 | 0.0008 | 0.71 | 0 |
| 16  | 53813367| FTO  | rs17817449 | G  | 0.39 | 5.02 | 5.56e−07 | 5.10e−08 | 0.0011 | 0.49 | 0 |
| 16  | 53620527| FTO  | rs9939609  | A  | 0.39 | 5.01 | 6.07e−07 | 8.53e−08 | 0.0011 | 0.75 | 0 |
| 16  | 53600954| FTO  | rs1421085  | C  | 0.39 | 4.65 | 0.0001 | 0.0011 | 0.0002 | 0.89 | 0 |
| 18  | 57673799| Near MC4R | rs12964056 | A  | 0.24 | −4.98 | 6.87e−07 | 0.0004 | 4.96e−06 | 0.95 | 0 |
| 3   | 42299870| CCK  | rs152472   | A  | 0.37 | −4.85 | 1.33e−06 | 0.0007 | 0.0002 | 0.89 | 0 |
| 4   | 142703670| ILLS | rs2009884 | T  | 0.15 | −3.17 | 0.0001 | 0.03 | 8.08e−05 | 0.19 | 0.41 |
| 2   | 142865291| LRP1B| rs7583748 | G  | 0.15 | −3.81 | 0.0001 | 0.09 | 0.02 | 0.74 | 0 |
| 2   | 644953 | TMEM18 | rs7561317 | A  | 0.17 | −3.17 | 0.0001 | 0.09 | 0.02 | 0.74 | 0 |

*The direction of all effect (weighted $z$ scores) are for the minor alleles. The $I^2$ inconsistency metric was null to small for all of the markers ($I^2 = 0–42\%$).

shown in Figure 2 (risk allele, A). There was 0.4 $z$-score-unit difference in mean of BMI-$z$ score between homozygotes with risk and non-risk genotype in our pediatric cohorts (Figure 2). We further subdivided all cohorts into two age strata of less than 5 and above 5 years old. In meta-analyses of both strata, the minor allele (A) was associated additively with a higher BMI (Figure 2).

Next, we performed imputation-based association followed by conditional analysis to identify independent association in the FTO locus. We identified 71 SNPs at the first intronic region of FTO that were significantly associated with BMI in European ancestry, all of which were of similar magnitude ($Z$ score $4.4 < Z < 5.26$, $p$-values between $10^{-7} < p < 10^{-6}$) and with high linkage disequilibrium (LD) with each other ($r^2 > 0.8$; Figure 3A). The haplotype boundaries and recombination rate in this ancestry are also shown in Figure 3. Indeed, the SNP rs8050136 was in proxy with other well-known variants in this region including rs9939609 ($r^2 = 0.98$), rs17817449 ($r^2 = 0.99$), rs1421085 ($r^2 = 0.95$) and all resided in the same haplotype and are associated with obesity in adults (4–9). Therefore, no independent effect has been identified.

Table 3 shows the summary of SNP results after regression analyses under additive and recessive models adjusted for age, gender, and PC. No significant difference was observed when the cohort site is included as another covariate (Data not shown). Of note, and only in this intron, more than 10 polymorphic indels were also associated with BMI-$z$, in particular A/AT at chr16:53822169 (MAF = 36%) and TTTC/T at chr16:53829962 (MAF = 35%; $p = 2.41 \times 10^{-7}$, $p = 8.09 \times 10^{-10}$, respectively). We noticed a subtle improvement of overall results using recessive model in our cohorts (Figure 3B; Table 3). In particular, another published SNP
rs1421085 generated the best result $z = 5.782, p_{	ext{Bonf}} = 8.21 \times 10^{-9}$ (Table 2).

Although BMI is a continuous trait, standard cut-offs were also used to assess the burden of increased body weight on health as a binary phenotype and estimate the effect size. When the tail distribution was considered ($>95\%$ as case and $<20\%$ as control), an OR of 1.61 ($95\% \text{ CI} = 1.31–1.97$, $p = 4.84 \times 10^{-5}$) was detected for the best surrogate marker rs8050136, adjusted by age, sex, and PC. The strongest effect size was observed using a recessive model for rs1421085 ($|OR| = 2.79$, $95\% \text{ CI} = 1.89–4.10$, $p = 1.85 \times 10^{-7}$).

When heterogeneity was allowed between cohorts, weaker signals were also obtained in loci, such as near the MC4R region (rs12964056, $p = 6.87 \times 10^{-7}$, $z = −4.98$), cholecystokinin CCK (rs8192472, $p = 1.33 \times 10^{-6}$, $z = −4.85$), Interleukin 15 (rs2099884, $p = 1.27 \times 10^{-7}$, $z = 4.34$), low density lipoprotein receptor-related protein 1B (LRP1B) (rs7583748, $p = 0.00013$, $z = −3.81$), and near transmembrane protein 18 (TMEM18; rs7561317, $p = 0.001$, $z = −3.17$), all of which have been previously reported to be associated with obesity or BMI (Table 3; 4–10). The imputation result for the MC4R region shows multiple association markers (Figure 3C).

We also performed imputation in additional regions of interest. Of note, one new locus that has not been previously reported to be associated with BMI passed the GWAS significance level when the best marker rs8050136 produced $|OR| = 2.90$ ($95\% \text{ CI} = 1.93–4.34$, $p = 9.03 \times 10^{-4}$) was obtained for this marker with MAF of 9% in cases vs. 3% in controls. Other unreported loci with suggestive associations to pediatric BMI ($10^{-7} < p < 10^{-3}$), include KCNH5 (rs10136789, $p = 4.62 \times 10^{-7}$, $z = 5.05$), APOL5 (rs2016586, $p = 3.26 \times 10^{-6}$, $z = −4.67$), LRRRC7 (rs10889850, $p = 1.77 \times 10^{-5}$, $z = −4.78$), and GALNT13 (rs12693973, $p = 1.65 \times 10^{-5}$, $z = −4.80$).

Finally, using ICD-9 diagnostic codes in our collections, we have also performed a Phewas study for the best identified markers in the FTO locus. In this approach, presence or absence of each diagnostic ICD-9 code was included as a binary phenotype, allelic associations were assessed between cases and controls, and the final results were corrected for multiple testing. By this approach, a negative association was detected between cases with hypertrophic cardiomyopathy or valvular structural heart disease and the FTO common risk alleles (ICD-9 codes = 424.0–424.3 and 425) in a subset of 81 cases and 2259 controls. This effect remained significant after 10,000 permutations ($p_{\text{perm}} = 0.0009$, $OR = 0.53$ ($95\% \text{ CI} = 0.37–0.77$)). Suggestive positive associations were also observed between the risk alleles and ICD-9 codes for impaired glucose tolerance test (ICD-9 = 790.2) and myopia (ICD-9 = 367.1; $p < 0.05$), however, this effect did not remain significant after permutation and correcting for multiple testing. It is noteworthy to mention that in our pediatric cohorts, the number of patients with diabetes related diagnostic codes (one of the BMI-related phenotype) was small and not sufficient for independent analysis.

**DISCUSSION**

In this study, we evaluated BMI in five European ancestry pediatriic cohorts with available EMR-linked genotyped data from the CCHMC–BCH eMERGE-II Network site. We successfully utilized anthropomorphic measures to calculate BMI-for-age percentile, derived BMI-z scores according to CDC growth charts, performed quantitative trait locus GWAS study and conducted meta analyses. The overall adjusted meta-analysis result of 2860 European samples produced the best signal in the 16q12 genomic region at the first intron of the FTO locus for a cluster of SNPs. The best typed marker rs8050136 produced $|OR| = 1.61$ ($95\% \text{ CI} = 1.31–1.965$, $p = 4.04 \times 10^{-4}$) was detected. Notably, this OR estimate in our data was relatively higher than previous adult studies (OR = 1.2, Loos, 2012). In any case, considering the effect size (or OR) at the range of 1.2 and the high MAF of the FTO loci (0.40) in Europeans, 700 samples were sufficient for us to achieve an optimum power of 0.8 with a type 1 error level of 0.05. In fact, genetic variants in the first intron of FTO present as the strongest BMI-associated GWAS locus in humans. Over the past few years, the association of the FTO locus has been repeatedly replicated, not only for BMI, but also for obesity risk, body fat percentage, waist circumference, and other obesity-related traits, in particular type II diabetes (Scuteri et al., 2007; Frayling et al., 2007). Recent association with dyslipidemia, hypertension, reduced brain volume, Alzheimer’s disease, and dementia has also been reported that could be confounded with obesity and vascular complications (Pausova et al., 2009; Keller et al., 2011).

The functional mechanism of FTO however is still elusive and is currently the subject of intense interest. It is an AlkB-like, Fe(II)-
and 2-oxoglutarate-dependent nucleic acid demethylase that has been shown to demethylate 3-methylthymine and 3-methyluracil in single-stranded DNA and RNA, respectively (Gerken et al., 2007). A link between FTO demethylase activity and increased fat mass was suggested by recent animal studies. Notably, homozygous mutant fto<sup>−/−</sup> mice show postnatal growth retardation and a significant reduction in adipose tissue and lean body mass, an observation that also was supported by the deleterious mutation Ile367Phe in mouse FTO protein with an impaired activity from a separate study (Church et al., 2009; Gao et al., 2010). In both studies, the leanness of fto-deficient mice seems to be the result of increased energy expenditure and systemic sympathetic
activation. Interpreting energy expenditure from these results, however, was challenging because of body composition differences and growth retardation. Overall, these experimental data suggest that inactivation of the FTO gene protects from obesity. On the other hand, mice globally overexpressing FTO are obese, hyperphagic, and exhibit normal energy expenditure when corrected for lean tissue mass (Church et al., 2010). Most human studies also report that obesity-predisposing FTO alleles are associated with increased food intake, but not energy expenditure (Speakman et al., 2008; Haupt et al., 2009). Although the detailed mechanisms are still not clear, this is suggested to be the result of increased expression of FTO in humans, due to cis-regulatory variation in intron 1 of this gene which is consistent with the mouse models, in which decreased levels of FTO cause a lean body habitus (Zabena et al., 2009; Berulava and Horsthemke, 2010). Moreover, a positive correlation of FTO gene expression with other adipocytokine gene expressions, including leptin, perilipin, and visfatin, has also been shown (Haupt et al., 2009).

Consistent with previous studies and because of high LD between FTO intron-1 markers in Europeans (Figure 2), conditional analyses didn’t produce independent effect; however, we observed a subtle improvement of FTO association using recessive models in comparison to additive models (Figure 1, Table 2). This effect was unique to the FTO locus, while using a dominant model was best near the MC4R region (Table 3). Indeed, between-study variations in regard to the optimal inheritance model of the FTO polymorphisms have been noticed previously. In one study, eight different meta-analysis results of BMI in women with polycystic ovary syndromes were systematically reviewed and the recessive model was found to fit best in half of them and the additive model worked best in the rest (Wójcickowski et al., 2012). In our pediatrics population in which we had an enrichment of earlier age of obesity with a prevalence rate of 28% (in recessive pattern) may be more relevant in comparison to the general population. Under this model, another intron-1 marker, rs4121085, produced the best result \( p_{\text{FDR}} = 8.21 \times 10^{-6}, z = -4.98 \), cholecystokinin CCK (rs8192472, \( p = 1.33 \times 10^{-7}, z = -4.85 \)), Interleukin 15 (rs2099884, \( p = 1.27 \times 10^{-7}, z = 4.34 \)), low density lipoprotein receptor–related protein 1B \[ LRP1B (rs758374, p = 0.00013, z = -3.81) \] and near transmembrane protein 18 (TMEM18; rs7561317, \( p = 0.001, z = -3.17 \)) (Table 2). Because of their lower effect on BMI and obesity risk (OR \( \sim 1.10 \), in adult meta-data) and lower allele frequency, the identification of these loci requires a quadrupling of the sample size in a random population. Finding all of these loci in our pediatric collections, despite limited sample size, indicate the enrichment of the genetic signal to noise ratio given the shorter amount of time that environment has had an effect.

Additionally, we have detected a new unreported signal at chromosome 3 (Co6A5) (best SNP is rs1542828, MAF of 5% \( p = 4.35 \times 10^{-5}, z = 5.89 \)). This marker produced an OR of 2.90 when the tail distributions (>95% as case and <20% as control, 386 cases, and 572 controls) was considered as binary phenotype. Considering this level of OR (2.90), even with MAF of 5%, 500 samples were sufficient for us to achieve the extraordinary power (0.99) with a type I error level of 0.05. This could be considered as one of the rare obesity risk loci that we expect to detect in these special cohorts. The u5-containing collagen VI (Co6A5, COL29A1), belongs to the class of collagens containing von Willebrand factor type A domains. These collagens form filaments with globular domains containing vWA motifs, which are involved in protein-ligand interactions for the organization of tissue architecture and clinical manifestations including severe intrauterine growth retardation, severe microcephaly, and death from infection before the age of three (Boissel et al., 2009). Of note, six out of eight cases had structural heart defect with cardiomyopathy. This would appear consistent with our unique observation of a negative association of obesity-predisposing FTO alleles with cardiomyopathy \( (p_{\text{FDR}} = 0.0009, \text{OR} = 0.53 \text{ [95% CI 0.37–0.77]}) \) that could result in lower expression of FTO; although further studies with larger sample sizes are necessary to confirm or refute this finding. In fact, to our knowledge, no SNP in intron 1 of FTO has been previously associated with any trait unrelated to BMI. Recently, an independent effect at the eighth intron of the FTO locus has been reported to be associated with melanoma (GenoMEL Consortium et al., 2013). The best marker, rs16953002, was replicated using 12,313 cases and 55,667 controls of European ancestry in a study conducted by the GenoMEL consortium (combined \( P = 3.6 \times 10^{-12}, \text{OR} = 1.16 \)). Notably, in their study, none of the BMI related SNPs in intron 1 were associated with melanoma. Similarly, in our collection, there was no effect observed at the eighth intron of the FTO gene with BMI \( (p = 0.54 \text{ for rs16953002, } r^2 < 0.01 \text{ with rs8050136}) \). This suggests independent functions and genetic risks for FTO that broaden the existing paradigm and identify distinct pathogenic effects (GenoMEL Consortium et al., 2013).

In this report, we have supported association in other previously reported BMI loci, in particular loci near the MC4R region (intra2964056, \( p = 6.87 \times 10^{-7}, z = -4.98 \)), cholecystokinin CCK (rs8192472, \( p = 1.33 \times 10^{-7}, z = -4.85 \)), Interleukin 15 (rs2099884, \( p = 1.27 \times 10^{-7}, z = 4.34 \)), low density lipoprotein receptor–related protein 1B \[ LRP1B (rs758374, p = 0.00013, z = -3.81) \] and near transmembrane protein 18 (TMEM18; rs7561317, \( p = 0.001, z = -3.17 \)) (Table 2). Because of their lower effect on BMI and obesity risk (OR \( \sim 1.10 \), in adult meta-data) and lower allele frequency, the identification of these loci requires a quadrupling of the sample size in a random population. Finding all of these loci in our pediatric collections, despite limited sample size, indicate the enrichment of the genetic signal to noise ratio given the shorter amount of time that environment has had an effect.
cell adhesion. Collagen VI is a major extracellular matrix (ECM) protein with a critical role in maintaining skeletal muscle functional integrity. It has been suggested that type VI collagen is a fibronic component that restricts adipose tissue expandability.

In humans, COL6A3 gene expression in adipose tissue was found to correlate with visceral adipose tissue mass and pro-inflammatory gene expression (Pauzart et al., 2009). Mutations in different families of this gene have also been associated with myopathy and muscular dystrophy. Recently, it has been shown that COL6A5 is involved in adhesion at myotendinous and dermal–epidermal junctions (Sahabzadeh et al., 2012). Different polymorphisms in or near this gene have been linked to atopic dermatitis and eczema, but with contradictory reports (Soderhall et al., 2007; Naumann et al., 2011). In our cohorts, the number of samples with Atopic dermatitis and related conditions (ICD-9: 691) was only 49 with a trend of association for published SNP rs7629719 (p = 0.14); adjusting the results based on presence or absence of atopy, didn’t have any effect on overall BMI associations. A larger sample size is necessary to further elucidate this coexistent condition and to determine whether COL6A5 has any role in obesity related conditions.

Four additional novel loci with homogeneous but suggestive associations (10^−3 < p < 10^−4) to childhood BMI were also identified in this study. These include KCNHS (rs10126789, p = 4.62 × 10^−3, z = 5.05), a voltage-gated potassium channel with various function in neurotransmitter regulation, hormone release, cardiac function, and cell volume; APOE (rs2015866, p = 3.26 × 10^−4, z = −4.67), a component of high-density lipoprotein with a potential role in lipid metabolism; LRRC7 or Denm (rs10888950, p = 1.77 × 10^−4, z = 4.78) a core component of post-synaptic densities and GALNT13 (rs12693973, p = 1.65 × 10^−3, z = 4.80) a member of the UDP-N-acetyl-alpha-D-galactosamime and a major enzyme responsible for the synthesis of O-glycan. Independent cohorts are necessary to confirm these preliminary suggestive findings and their importance in childhood obesity.

Despite the limitations of using an EMR-derived data set for analysis of secondary phenotypes including errors in data extractions, discordant time of sampling, and underlying coexistent conditions of our pediatric cohorts, we demonstrate that a strong signal, larger than seen in adult populations is detectable. We have removed all inconsistent data and outliers to the best of our ability. We have also excluded infants and those less than 2 years old because of the complexity of growth chart pattern and many potential maternal effects on infants from perinatal periods. In addition we assessed the distribution of BMI-z in the whole population with a large sample size as a quantitative trait rather than attempting to identify limited cases and controls. From the statistical standpoint, quantitative traits usually are preferred in meta-GWAS studies because they improve power to detect a genetic effect and often have a more interpretable outcome (Bush and Moore, 2012). Furthermore, BMI is a highly heritable trait in humans and, as mentioned above, up to 70% of the inter-individual variation in obesity can be attributable to genetic factors per se (Maes et al., 1997), therefore, given the strong genetic confirmation described here, indeed, we managed to re-interpret the genotyping data collected for the analyses of another phenotype and successfully find association between the new phenotype and genotypic data.

In summary, using the EMR-linked genotyped data, we have confirmed association of several previously known BMI loci, in particular with the FTO gene (OR of 1.61 (95% CI = 1.31–1.97)). Our data also support the importance of variants at the FTO locus in childhood obesity and with saturation of an early age, these data point to a closer functional variant in this locus.

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