

Early Ethnic Difference in Insulin-Like Growth Factor-1 Is Associated with African Genetic Admixture

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ABSTRACT

IGF-1 is a growth-promoting hormone. Numerous studies have reported higher systemic concentrations of IGF-1 among African Americans (AA) compared with European Americans (EA) before puberty. We conducted this cross-sectional analysis to determine whether African ancestral genetic background, dietary factors, energy expenditure, adiposity, and socioeconomic status contribute to this difference. Children were prepubertal, AA and EA males and females. Genetic admixture was assessed from ~20 ancestry informative genetic markers. Body composition was determined by dual-energy x-ray absorptiometry; intake of energy, carbohydrate, protein, and fat by 24-h dietary recall; activity-related energy expenditure by doubly labeled water and indirect calorimetry; and socioeconomic status (SES) according to the Hollingshead scale. IGF-1 and IGF binding protein-3 (IGFBP-3) were measured using immunoradiometric assays. AA children had significantly greater IGF-1 compared with EA children ($p < 0.01$). In addition, AA children had lower

SES and greater protein intake relative to EAs ($p < 0.05$ for both). Multiple linear regression analysis revealed that the only significant independent predictors of IGF-1 were IGFBP-3 and African admixture ($p < 0.01$ for both). Thus, our data suggest that the greater IGF-1 of AA relative to EA children could have a genetic basis. (*Pediatr Res* 58: 850–854, 2005)

Abbreviations

AA, African American
AEE, activity-related energy expenditure
AFADM, African admixture
EA, European American
GH, growth hormone
IGF-1, insulin-like growth factor-1
IGFBP-3, insulin-like growth factor binding protein-3
SES, socioeconomic status

IGF-1 is a potent anabolic peptide that is present in the systemic circulation and is expressed in most body tissues. Thus, the peptide has both endocrine and autocrine/paracrine actions (1). Its systemic and tissue concentrations are regulated by numerous factors including growth hormone (GH) and an array of binding proteins. IGFBP-3 is the dominant regulator of systemic IGF-1 bioavailability, binding the majority of the circulating peptide (1). Because of its potency and ubiquity within the body, IGF-1 is considered an important endogenous

growth-promoting agent. However, IGF-1 is also associated with numerous disorders, including cancer (2,3).

Several studies have shown that AA children have higher systemic concentrations of IGF-1 compared with EA children (4–6). Most studies suggest that this difference does not track into adulthood (7,8), but some data do indicate higher concentrations in AA compared with EA adults (9). The physiologic basis for higher IGF-1 among AAs is not known. An obvious explanation is that AAs secrete more GH, however, it has been suggested that GH secretion does not differ between AAs and EAs in childhood (4,10).

Ethnic differences in physiologic variables could be related to environmental or genetic factors. Epidemiologic and clinical studies have detailed numerous environmental factors associated with systemic IGF-1 concentrations (11–13). Higher IGF-1 is associated with higher intake of total energy, carbohydrate, fat, and protein (11,12). Adiposity and physical activity also may influence IGF-1 (13,14). It is not known whether these variables contribute to the ethnic difference in IGF-1 during childhood. Conversely, ethnic differences in IGF-1 may

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have a genetic basis. The recently developed ancestral genetic admixture analysis can be used to assess genetic influences on physiologic variables that differ by ethnicity (15). Modern-day North Americans represent a population created by intermixing of European, African, and Native American populations. Hence, alleles from previously insulate populations were brought together in combinations that now encompass the gene pools of the various ethnic groups residing in North America. The relative contribution of parental population (*e.g.* African) to an individual's genome can be estimated using specific ancestry informative DNA markers that differ in frequency among the respective parental populations. Hence, assigning individuals an estimate of ancestral admixture, such as percentage AFADM, can help to delineate genetic from environmental influences on phenotypes that differ by ethnicity (15).

Therefore, we undertook the present analysis to determine whether diet, energy expenditure, SES, or adiposity contributed to the ethnic difference in systemic IGF-1 in childhood. In addition, to provide further insight into the influence of genetic factors on the ethnic difference in IGF-1, we assessed whether AFADM was an independent predictor of IGF-1. We hypothesized that ethnic differences in the nongenetic variables and/or adiposity would explain the higher IGF-1 of AA children. However, we found that the only independent predictors of IGF-1 were IGFBP-3 and AFADM, indicating a possible genetic basis for early ethnic differences in IGF-1.

METHODS

Sample. Data for our present analysis were derived from an ongoing longitudinal study of body fat distribution and disease risk factors in children and adolescents. Complete measures of dietary intake, energy expenditure, and body composition were available on 55 prepubertal children. Pubertal stage was determined by a pediatrician, according to the criteria of Tanner and Marshall (16,17). All children were prepubertal AAs (13 boys and 13 girls) and EAs (16 boys and 13 girls). Child ethnicity was assigned based on parental and grandparental ethnicity, *i.e.* both parents and all four grandparents had to be AA or EA for a child to be considered AA or EA, respectively. Children were recruited by word-of-mouth and newspaper advertisements. All participants recruited into this study were required to be greater than 4 y of age and free of medications or disorders known to affect body composition, metabolism, or physical activity. The nature, purpose, and possible risks of the study were fully explained to parents before consent was obtained. All testing was undertaken at the University of Alabama at Birmingham (UAB) General Clinical Research Center (GCRC) and the Department of Nutrition Sciences. The study was approved by the Institutional Review Board for Human use at UAB. Written informed consent was obtained from parents and children before testing was initiated.

Energy expenditure assessment. Total free-living energy expenditure was determined over 14 d with the doubly labeled water technique as previously described (18). This method requires determination of the washout kinetics of isotopes of hydrogen ($^2\text{H}_2$) and oxygen ($^{18}\text{O}_2$), which are administered at baseline. CO_2 production, an indirect measure of metabolic rate, is then determined as the difference in the disappearance rate of $^2\text{H}_2$ and $^{18}\text{O}_2$. This can be achieved because the rate of $^{18}\text{O}_2$ disappearance is influenced by both fluid loss and CO_2 production, whereas the rate of $^2\text{H}_2$ disappearance is influenced predominantly by fluid loss. In brief, four-timed urine samples were collected after oral dosing with doubly labeled water, two samples were obtained in the morning after dosing and two in the morning 14 d later, with a loading dose of 0.15g of H_2^{18}O and 0.12g of $^2\text{H}_2\text{O}$ per kilogram of body weight. Samples were analyzed in triplicate for H_2^{18}O and $^2\text{H}_2\text{O}$ by isotope-ratio mass spectrometry as previously described (18). Carbon dioxide production was then determined using the R2 equation of Speakman *et al.* (19), assuming a fixed dilution space ratio of 1.0427, and energy expenditure was calculated using Equation 12 of de Weir, assuming the mean value of the dietary quotient was 0.9 (20). This mean value for the food quotient of the subjects' diet was previously estimated from 24-h dietary recalls in this group.

This protocol has a theoretical error of <5% for the measurement of total energy expenditure (18).

AEE was estimated as the difference between total energy expenditure and resting energy expenditure (REE), adjusted for the thermic effect of food [AEE = $(0.9 * \text{TEE}) - \text{REE}$]. REE was measured by indirect calorimetry using a Deltatrac Metabolic Monitor (Sensormedics Corp, Yorba Linda, CA). Subjects were awakened at 0500 h, after an overnight stay at the GCRC, and instructed to lie in the supine position and remain motionless while an adult-sized canopy hood was used to collect expired gases. After a 10-min equilibration, oxygen consumption and carbon dioxide production data were collected continuously for 20 min. REE was calculated using the equation of de Weir (20). In children, measurement of REE with the Deltatrac has been shown to be highly reliable (CV of 5.8% for six repeat measurements on six different days) and a single measurement is considered adequate to determine REE in children (21).

Blood sampling and hormone analysis. Fasting blood samples were obtained after measurement of REE. Serum samples were analyzed in duplicate for IGF-1 using a standard immunoradiometric assay (Diagnostic Systems Laboratories, Webster, TX; intra-assay CV was 3.7%, inter-assay CV was 7.3%) and in duplicate for IGFBP-3 using an immunoradiometric assay (Diagnostic Systems Laboratories; intra-assay CV was 4.0%, inter-assay CV was 7.7%). Serum estradiol was determined using a double-antibody RIA (Diagnostic Systems Laboratories; intra-assay CV was 3.6%, inter-assay CV was 5.2%). Serum testosterone was assessed using a solid phase RIA (Diagnostic Systems Laboratories; intra-assay CV was 2.7%, inter-assay CV was 8.6%). Using these assays, the minimum detectable values of estradiol and testosterone were 4.2 pg/mL and 11.8 ng/dL, respectively.

Body composition assessment. Body composition (total body fat mass and nonbone lean tissue mass) was measured by dual-energy x-ray absorptiometry (DXA) using a Lunar DPX-L densitometer (Lunar Radiation Corp., Madison, WI). Subjects were scanned in light clothing, while lying flat on their backs with arms at their sides. DXA scans were performed and analyzed with pediatric software version 1.5e. DXA has been found to be highly reliable for body composition assessment in children; in our laboratory, the CV for repeated measures of total body fat mass was 6.55%. Height was measured to the nearest centimeter using a stadiometer, and weight was measured on an electronic scale while children wore light clothing.

Assessment of diet and socioeconomic status. Habitual intake of energy, protein, and fat was estimated from the average of three 24-h dietary recalls (one of which included a weekend day) using the multiple-pass technique in the presence of one or both parents. Each dietary recall was administered by a trained technician. The protocol was as follows: 1) a list of foods consumed in the previous 24 h was obtained; 2) details regarding the specific type of food and portion sizes consumed at each meal was determined; 3) lastly, a review was conducted to ensure that no food items consumed in the previous 24 h were excluded. The 24-h recall method for assessing energy intake has been validated relative to the doubly labeled water method for assessment of energy expenditure in children. Comparisons of the mean energy intake and energy expenditure (assessed by doubly labeled water) did not differ significantly (22). Data were then entered into the Food Intake Analysis System (FIAS version 3.0, University of Texas, 1996) for determination macronutrient composition. The FIAS program is based on the US Department of Agriculture Nationwide Food Consumption Survey, Continuing Survey of Food Intakes by Individuals, and allows for the addition of new foods, recipe modification, and the selection of typical food portion sizes. The FIAS analysis provided average measures of total energy intake (kcal/d), protein intake (g/d), carbohydrate intake (g/d), saturated fat intake (g/d), and total fat intake (g/d). SES was determined according to the Hollingshead index of social status (23). This scale combined the education level and occupational prestige for the number of working parents in each child's family, accounting for sex and marital status. Social class scores using this scale range from 8 to 66; higher values represent a higher SES.

Determination of genetic admixture. Genotyping of study participants was carried out at the Pennsylvania State University. Genotyping of single nucleotide polymorphisms was carried out using agarose gel electrophoresis and melting curve analysis, where a fluorescent-labeled sequence-specific probe (SYBR Green I) is added during PCR that creates a heating-dependent melting curve for each polymorphism. DNA strand dissociation causes loss of fluorescence, which is a function of the PCR product length, its sequence composition, and its GC content. Melting curve analysis is described in detail by Akey *et al.* (24). The ancestry informative markers used, their chromosomal and centimorgan location, and their ability to discriminate among parental populations (based on the allelic differences between European and African parental populations) are detailed in Table 1 and are described elsewhere (25). Further information about these markers is available through dbSNP (<http://www.ncbi.nlm.nih.gov/projects/SNP/>) using handle PSA-ANTH. All genetic data were transformed into a single estimate of genetic admixture using the

Table 1. Ancestry informative DNA markers used in the determination of African genetic admixture, their location, and frequency difference (δ) among parental populations

| Marker | Chromosomal location | cM | δ (AA vs. EA ¹) |
|--------------|----------------------|-------|------------------------------------|
| MID 575 | 1p34.3 | ~64 | 0.130 |
| MID 187 | 1p34.1 | ~75 | 0.370 |
| FY-Null | 1q23.2 | ~165 | 0.999 |
| AT3 | 1q25.1 | ~191 | 0.575 |
| WI-11392 | 1q42.2 | ~252 | 0.444 |
| WI-18657 | 2p16.1 | ~79 | 0.536 |
| WI-11153 | 3p12.3 | ~106 | 0.652 |
| GC*1F | 4q13.3 | 79 | 0.697 |
| GC*1S | 4q13.3 | 79 | 0.538 |
| SGC30055 | 5q23.1 | ~120 | 0.457 |
| CYP3A4 | 7q22.1 | ~111 | 0.761 |
| LPL | 8p21.3 | ~398 | 0.479 |
| D11S429 | 11q11 | ~70.9 | 0.429 |
| DRD2-Taq1"D" | 11q23.1 | ~105 | 0.535 |
| APOA1 | 11q23.3 | ~113 | 0.505 |
| GNB3 | 12p13.31 | ~15 | 0.463 |
| OCA2 | 15q13.1 | ~16 | 0.631 |
| MC1R314 | 16q24.3 | ~133 | 0.350 |
| WI-14867 | 17p13.2 | ~10 | 0.448 |
| WI-7423 | 17p12 | ~16 | 0.476 |
| Sb19.3 | 19p13.11 | ~49 | 0.488 |
| MID154 | 20q11.22 | ~50 | 0.444 |

¹ (δ) Allelic frequency difference between West African and European parental populations; calculated as the frequency of marker allele in West African populations minus the frequency in European populations.

maximum likelihood approach (26) such that the likelihood that a particular multilocus genotype is from each of 100 different combinations of the two parental populations is calculated as the probability of the genotype given the allele frequencies in each parental population. The parental population combination where the specific multilocus genotype for an individual has the highest probability is thus the most likely and represents the admixture estimate for that individual.

Statistics. Two-way ANOVA was used to assess potential effects of ethnicity and gender on the variables analyzed. Relations between IGF-1 and dietary, activity, and other outcome variables were assessed using Pearson correlation coefficients. Independent associations of genetic, lifestyle, and physiologic variables with IGF-1 were assessed using multiple linear regression analysis. All data were log transformed to correct for skewed distribu-

tions. Statistical significance was set at $p < 0.05$ for all analyses. Data were analyzed using SPSS for Windows version 10.0 (SPSS Inc., Chicago, IL).

RESULTS

IGF-1 concentrations were significantly higher in AA children compared with EA children (Table 2). Girls had significantly higher IGF-1 concentrations than boys; there was no interaction between gender and ethnicity (Table 2). Girls also had significantly higher IGFBP-3 concentrations than boys; no ethnic difference in IGFBP-3 was noted. SES was significantly lower in AA children and total protein intake was higher. There were no ethnic or gender differences in age, height, weight, fat mass, energy intake, fat intake, or AEE (Table 2). Two boys (one AA, one EA) had serum testosterone concentrations above the detectable limit and one AA girl had a serum estradiol concentration above the detectable limit. There were no ethnic or gender differences in concentrations of sex hormones (Table 2) and all children were prepubertal (Tanner stage one).

Pearson correlation coefficients for IGF-1 are presented in Table 3. SES was negatively associated with IGF-1. AFADM, IGFBP-3, fat mass, and AEE were significantly and positively correlated with IGF-1. There were no significant correlations between IGF-1 and the dietary intake variables.

Results from the multiple regression analysis for the dependent variable IGF-1 are presented in Table 4. Variables that were significantly correlated with IGF-1 or were significantly different between ethnicities were added to the final multiple regression model. Hence, independent variables in the analysis were AFADM, fat mass, protein intake, SES, AEE, and IGFBP-3. The only variables independently associated with IGF-1 were AFADM ($p < 0.01$) and IGFBP-3 ($p < 0.01$); the R^2 value for the model was 0.59. Addition of gender to the model did not affect the independent association of IGF-1 with AFADM (data not shown).

DISCUSSION

We tested the hypothesis that the higher concentrations of systemic IGF-1 in AA children compared with EA children were related to environmental and body composition variables known to be associated with IGF-1. AA and EA children did not differ significantly by age, body weight or height, pubertal stage, or sex hormone concentrations. Contrary to our hypothesis, we found that neither diet, physical activity, SES, nor adiposity were related to the higher IGF-1 of AA children. Results indicated that AFADM and IGFBP-3 were the only independent predictors of IGF-1 in our sample. IGFBP-3 has a strong influence on systemic IGF-1 because it binds the ligand

Table 2. Subject characteristics* Mean \pm SD

| | EA (n = 29) | AA (n = 26) | Ethnicity [†] | Gender [†] | Ethnicity \times gender [†] |
|----------------------|------------------|-------------------|------------------------|---------------------|--|
| Age (yrs.) | 7.9 \pm 2.0 | 8.3 \pm 1.4 | NS | NS | NS |
| Weight (kg) | 32.3 \pm 7.4 | 36.6 \pm 12.4 | NS | NS | NS |
| Height (cm) | 131.5 \pm 8.9 | 134.8 \pm 9.7 | NS | NS | NS |
| Fat mass (kg) | 8.9 \pm 4.5 | 11.1 \pm 8.1 | NS | NS | NS |
| IGF-1 (ng/mL)** | 190.8 \pm 79.3 | 258.2 \pm 100.2 | <0.01 | <0.05 | NS |
| IGFBP3 (ng/mL)** | 3791 \pm 794 | 3939 \pm 621 | NS | <0.05 | NS |
| Testosterone (ng/dL) | 11.87 \pm 0.47 | 11.84 \pm 0.23 | NS | NS | NS |
| Estradiol (pg/mL) | 4.20 \pm 0.00 | 4.30 \pm 0.06 | NS | NS | NS |
| SES | 53.2 \pm 7.5 | 33.3 \pm 13.2 | <0.001 | NS | NS |
| Energy (Kcal/d) | 1854 \pm 501 | 1858 \pm 525 | NS | NS | NS |
| Carbohydrate (g/d) | 260.4 \pm 83.2 | 225.5 \pm 80.7 | NS | NS | NS |
| Protein (g/d) | 58.9 \pm 19.9 | 76.9 \pm 37.7 | <0.05 | NS | NS |
| Fat (g/d) | 66.6 \pm 23.2 | 74.9 \pm 26.1 | NS | NS | NS |
| Saturated fat (g/d) | 23.3 \pm 7.7 | 26.9 \pm 13.6 | NS | NS | NS |
| AEE (Kcal/d) | 667.5 \pm 1152 | 663.6 \pm 382 | NS | NS | NS |
| AFADM (%) | 3.4 \pm 6.4 | 84.8 \pm 11.0 | <0.001 | NS | NS |

* Statistical comparisons were conducted with log transformed data to correct for skewed distributions; untransformed mean \pm SD are presented for ease of interpretation.

[†] P-values for 2-way ANOVA for ethnicity, gender, and the interaction between ethnicity and gender. NS = not significant ($P > 0.05$).

** To convert metric units (ng/mL) to SI units (nmol/l) for IGF-1 multiply by 0.131, for IGFBP-3 multiply by 0.035.

Table 3. Pearson correlation coefficients for the relation between IGF-1 and dietary, lifestyle, and physiologic variables*

| | <i>r</i> | <i>P</i> |
|---------------|----------|----------|
| Age | 0.12 | 0.37 |
| IGFBP-3 | 0.70 | <0.001 |
| SES | -0.29 | <0.05 |
| Energy | 0.02 | 0.89 |
| Carbohydrate | -0.09 | 0.52 |
| Protein | 0.09 | 0.49 |
| Fat | 0.14 | 0.31 |
| Saturated fat | 0.02 | 0.91 |
| Fat mass | 0.35 | <0.01 |
| AEE | 0.27 | <0.05 |
| AFADM | 0.35 | <0.01 |

* Analyses were conducted with log-transformed data to correct for skewed distributions.

Table 4. Independent predictors of serum total IGF-1 from multiple linear regression analysis*, †, variables excluded from model are described below

| | $\beta \pm \text{SEE}$ | <i>P</i> |
|-----------|------------------------|----------|
| Intercept | -3.47 \pm 0.72 | <0.001 |
| AFADM | 0.014 \pm 0.00 | 0.001 |
| IGFBP-3 | 1.59 \pm 0.20 | <0.001 |

* Analyses were conducted with log-transformed data to correct for skewed distributions.

R² for the model = 0.578. Variables excluded from the model at *P* > 0.05: SES (*P* = 0.38); fat mass (*P* = 0.62); protein (*P* = 0.90); AEE (*P* = 0.47).

† Gender did not influence the independent association of AFADM with IGF-1 (data not shown).

and increases its half-life. The association of AFADM with systemic IGF-1 suggests that higher IGF-1 in AAs relative to EAs may have a genetic basis.

Associations between disease and elevated concentrations of IGF-1 have been documented in adults (2,3). Furthermore, indices of childhood growth such as leg length have been linked to the development of hormone-related cancers such as those of the prostate and breast (27). It has been suggested that growth-promoting factors such as IGF-1 could mediate the link between growth and later cancer risk (28). The prevalence of prostate cancer is known to be higher in AA men (29), and AA women are more likely to have a more advanced breast tumors compared with EA women (30). High growth rates during childhood are also associated with later development of obesity and cardiovascular disease risk profiles (31,32). Obesity and cardiovascular-related morbidity such as stroke have higher prevalence rates in AAs compared with EAs (33). Whether IGF-1 differences in during childhood directly or indirectly contribute to ethnic differences in disease risk is unknown.

Systemic IGF-1 may be affected by diet and physical activity (14). The mechanism underlying the relation between physical activity and IGF-1 is not known but may be related to effects on energy and nitrogen balance, or on IGFBP regulation (14). A low-fat diet combined with a daily exercise intervention has been shown to decrease IGF-1 independent of IGFBP-3 concentrations in older men (13). Studies have shown that exercise and physical activity either have no association with, or are associated with both higher and lower IGF-1 (13,34–36). Although we found a positive association between IGF-1 and

AEE, our data suggest that differences in energy expenditure are not responsible for the ethnic difference in IGF-1.

Epidemiologic studies have demonstrated that higher energy, protein, and fat intakes are associated with higher IGF-1 (11,12,37), however, in one study dietary factors were not associated with IGF-1 in children (38). Dietary influences on IGF-1 production could be direct, or indirect, *via* macronutrient influences on GH secretion or IGFBP regulation; the actual mechanism(s) is unknown (14,36). We did not find that differences in total energy, protein, or fat intake were related to the higher IGF-1 of AA relative to EA children.

Adiposity has also been associated with IGF-1 in adults and children (14,39), and fat mass was positively correlated with IGF-1 in our study. However, we did not find fat mass to have an influence on the ethnic difference in IGF-1. Our observations suggest that physical activity, macronutrient intake, SES, and adiposity are not responsible for the ethnic difference in systemic IGF-1.

Measures of AFADM yield some insight into the influence of genetic *versus* environmental factors on ethnic differences in physiologic outcomes. In this study, AFADM independently predicted systemic IGF-1, suggesting that the higher IGF-1 of AAs compared with EAs may be attributable to genetic differences. Polymorphisms in several genes have been associated with systemic IGF-1 (9). Polymorphisms in the promoter region of the IGF-1 gene and in the promoter region of the gene for the cytochrome P450 3A4 enzyme have been shown to account for the higher IGF-1 of premenopausal AA women relative to EA women, however, this effect was only observed in those using oral contraceptives (9). It is not known whether any of these polymorphisms explain the ethnic differences in IGF-1 during childhood.

Differences in maturation rate could be responsible for a portion of the ethnic difference in IGF-1. AA children, particularly girls, enter into puberty on average 1 y earlier than EA children (40). IGF-1 concentrations rise in response to rising GH and sex-steroid hormone concentrations at the beginning of puberty. Hence, it is possible that the higher concentrations of IGF-1 in the prepubertal AA compared with EA children of this study reflected their greater proximity to puberty. Although the two groups did not differ with respect to concentrations of sex hormones, subtle differences in hormone concentrations may not have been detected.

Conversely, as GH and/or IGF-1 may play a role in the timing of puberty onset; higher IGF-1 among AA children may contribute to their earlier maturation. This theory is supported by observations that, relative to boys, girls have both a higher concentration of IGF-1 before puberty (38), and an earlier age of pubertal onset. In agreement with this previous observation, the girls in the present study had higher IGF-1 than the boys. Whether higher IGF-1 among AA children is a cause or consequence of earlier puberty requires further investigation.

There are several limitations to our study. These include sample size, and questionnaire-based measures of macronutrient intake and SES. Given that our study was cross-sectional in design and our sample size was limited to *n* = 54, our findings of the absence of an environmental component in the ethnic difference in IGF-1 should be considered tentative until these

findings can be replicated in a larger sample. Indeed, unmeasured environmental parameters may also have contributed to the higher IGF-1 of AA. In addition, because questionnaire and other recall-based assessments may have limited accuracy, the contribution of these variables to the higher IGF-1 of AAs relative to EAs cannot be completely ruled out. Furthermore, since IGF-1 concentrations are strongly influenced by GH, the higher IGF-1 of AAs could be related to higher GH secretion or action. Although GH secretion has not been shown to be greater in AA children compared with EA children, the sample studied comprised only a small number of males (10); therefore, higher GH cannot be completely discounted as the cause of the higher IGF-1 concentrations. Unfortunately, robust analysis of GH secretion, which requires frequent blood sampling throughout the day, was not available in our study. Further studies with larger samples sizes and accurate measures of GH release are needed to clarify this issue.

In conclusion, we found an independent association of AFADM with IGF-1. Hence, this ethnic difference could be related to genetic factors, perhaps *via* polymorphisms of genes directly regulating the GH/IGF system or regulating maturation rate. However, further work will be required to conclusively test these hypotheses. The effect of higher childhood IGF-1 concentrations on disease risk in this population remains to be determined.

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