Sleep Duration and Obesity in Children: Is the Association Dependent on Age and Choice of the Outcome Parameter?

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Study Objectives: To assess the association between sleep duration in children and different markers of body fat by age and weight status. **Design:** Nation-wide health survey. Measurement of BMI and body fat percentage (KFA) calculated from weight, height, skin fold thickness, age, and sex. Sleep duration and potential confounding variables were assessed in a parent questionnaire.

Setting: N/A

Participants: 7767 German resident children from 3 to 10 years of age.

Interventions: N/A

Measurements and Results: Prolongation of sleep duration from the lowest to the highest percentile accounted for a similar mean decrease in BMI (-0.235, 95%-Cl -0.321; -0.149) and KFA (-0.182, 95% Cl -0.271; -0.092) z-scores. The given association is adjusted for con-

A GROWING BODY OF EVIDENCE THAT SLEEP DURA-TION AFFECTS WEIGHT STATUS IN CHILDREN HAS BEEN SUMMARIZED IN 3 RECENT REVIEWS.¹⁻³ A metaanalysis by Chen and colleagues calculated a 9% decrease in overweight/obesity risk for a one-hour increase in sleep duration in children up to 10 years of age.

There is still considerable variability in the effects reported by different individual studies, which could be explained by different ages included. This issue has never been systematically assessed in publications the authors are aware of. The majority of studies use measures of overweight/obesity risk, while BMI is reported less frequently. The changes in BMI distributions have merely been addressed, although they might have implications for choosing the appropriate outcome measure and are interesting for physiological and epidemiological reasons. Furthermore, there is limited evidence on whether sleep duration is linked to body fat in children.⁴⁻⁶ A more detailed analysis of epidemiological data with respect to these issues may have implications for further research regarding underlying physiological mechanisms.

The aims of this study were

1) to describe the sleep duration typical for children 3 to 10 years of age in a nationally representative sample

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Address correspondence to: Otmar Bayer, MD, MPH, Institute for Social Paediatrics and Adolescent Medicine, Ludwig-Maximilians University, Heiglhofstr. 63, 81377 Munich; Tel: +49-89-71009-366; Fax: +49-89-71009-315; E-mail: Otmar.Bayer@Irz.uni-muenchen.de founding variables and did not show a systematic age dependency. The greatest effects of sleep duration were seen for the upper tails of the BMI and KFA distributions, which were about four as high as the lower tails.

Conclusions: The association between sleep duration and weight status is of similar size through ages 3 to 10 years. The sleep-associated changes in BMI are likely to be a consequence of higher body fat and primarily affect children whose BMI or KFA is already elevated. These findings favor hormonal pathways nurturing adipose tissue playing a key role in the underlying physiological mechanisms.

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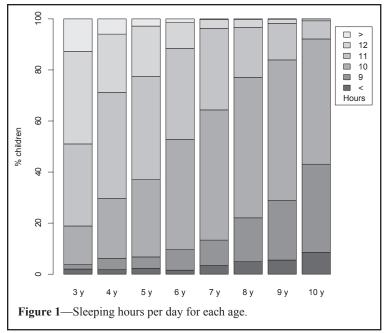
- 2) to assess whether the effects of sleep duration on body mass are age dependent
- 3) to investigate if these effects are also seen on body fat (KFA) as outcome
- 4) to analyze how the distribution of BMI and KFA is changed in relation to sleep duration

SUBJECTS AND METHODS

Data Collection

The German Health Interview and Examination Survey for Children and Adolescents (KiGGS) was conducted from May 2003 to May 2006 by the Robert Koch Institute and surveyed 17,641 children and adolescents from 0 to 17 years of age. Data were collected with the aim of establishing a nationally representative sample.⁷ Different age-appropriate questionnaires were used, and the present study included children from 3 to 10 years of age, for whom information on duration of sleep was collected in a uniform manner.

Participants were invited to one of 167 local study centers, where they filled in questionnaires followed by a physical examination. The local study teams were thoroughly trained and each consisted of 5 members, led by a physician experienced in pediatrics.⁸ Children's height (without shoes) was measured with an accuracy of 0.1 cm, using a portable Harpenden stadiometer (Holtain Ltd., Crymych, UK). Body weight (while wearing underwear) was measured with an accuracy of 0.1 kg with a calibrated electronic scale (SECA, Birmingham, UK). For non-German families with poor command of the German language, questionnaires in the native languages were provided. Sleep duration was assessed in the self-administered parent



questionnaire asking for the average hours of sleep per day (discrete value, allowing no decimal places after the comma).

Data Analysis

BMI was modelled as an age and gender-specific z-score obtained from the sample. Overweight and obesity were defined according to the International Obesity Taskforce (IOTF)⁹. This definition is based on centile curves drawn to pass through the widely used cutoff points of 25 and 30 kg/m² at age 18. The data for these curves were obtained from 6 large nationally representative cross-sectional growth studies of international origin. KFA (in %) was computed from the sum of skinfold thickness at the back and triceps region using Slaughter's formula.¹⁰ As with BMI, it was transformed to a z-score.

Sleep needs vary with age.^{11,12} As can be seen from Figure 1, it is hardly possibly to find cutpoints for sleep duration comparable through all age groups. For example, tertiles could be well applied in the 4-year-old children but do not fit in the 6-year-old children. To facilitate modelling we constructed a common measure sleep duration percentile (SDP): For each age, the cumulative frequency for the ordered discrete values of hours of sleep were computed. 11 hours of sleep in a 3-year-old child can be assigned to a cumulative frequency of 0.51, indicating, that such a child sleeps less than 1 - 0.51 = 49% and more than 18.9% (see leftmost bar in Fig. 1) of the population of his or her age. The real value for hours of sleep in a child given the discrete value of 11 h may vary between 10.5 and 11.5 h. In consequence, SDP was finally computed as the center of the adjacent cumulative frequencies, resulting in (0.189 + 0.510)/2 = 0.349 for the given example. Thus, SDP can be interpreted as a percentile, identfying the proportion of children with fewer hours of sleep. It should be re-emphasized that sleep duration was originally collected as discrete values (see end of Data collection), and the transformation does not result into a loss of precision.

To model the association between sleep and body mass as precisely as possible, we screened a list of 24 potentially confounding variables, concerning parental BMI, social class, township size, smoking during pregnancy, breastfeeding, birth, leisure time sport activity, and electronic media consumption. Among variables of similar content (e.g., socioeconomic status (SES) of father vs. SES of mother, or exclusively breastfed until 4th vs. 6th month) the variable was selected whose inclusion in the model led to the greatest change in estimate for the regression parameter of SDP. Further selection criteria were completeness and additional variance explained. The variables finally included in our multivariable model are SES of father, classified as lower, middle (reference), or upper class according to Winkler's index¹³; BMI of mother (or father if BMI of mother was not available); smoking during pregnancy, categorized as "never" (reference), "yes, from time to time" or "yes, regularly"; exclusively breastfed until 4 months; birth weight in grams; and electronic media consumption classified as low (reference), intermediate, or high.

The variable electronic media consumption was used as provided by the Robert Koch Institute, which for ages 3–10 was compiled as follows: Hours watching TV and hours playing computer games were asked separately for

hours playing computer games were asked separately for weekdays and weekends by 4 multiple-choice questions, each allowing 1 of 5 answers, between never and > 4 h per day. The answers were assigned to values between zero and 5 hours and then added up over the 4 questions to estimate the electronic media consumption in hours per week. Finally, these were classified as low, intermediate, or high, using approximate tertiles for each age as cutoff values.

Statistical Methods

To provide adjusted effect estimates for sleep duration on zscores for BMI and KFA, multivariable linear regression models were applied. Accordingly, adjusted effect estimates for sleep duration on overweight and obesity were calculated applying multivariable logistic regression. The correlations between potential confounders and the explanatory/outcome variables were quantified using Pearson r. While linear (ordinary least squares) regression models the conditional mean of an outcome variable Y for its whole distribution, quantile regression allows to model the conditional mean at any part of the distribution in relation to the exposure of interest. This method is warranted, if the effects of the exposure X are not homogeneous among the distribution of Y.¹⁴ This method was applied to identify potential causes for possible differences in the observed effects when using different outcome measures (metric vs. categorical outcomes).

Statistical analysis was done using R version 2.6.2 on Linux. $^{\rm 15-18}$

RESULTS

The overall response in the KiGGS study was 66.6% (49:51 female-to-male)⁷ resulting in 17,641 children participating. Basic health-related variables were similar among responders and non-responders; in particular BMI and "smoking mother" showed no significant differences. Participation was homogeneous with regard to age (ranging from 63% to 70%) and gender (67% vs. 66% female vs. male). On average, mothers of responders completed more education than mothers of non-

Table 1—Number of Observations, Proportion Male, Overweight, Obese, and Mean (± std) BMI and KFA for the Study Population

age (years)	n	% male	% overweight	% obese	BMI girls	KFA girls
3	934	51.2	8.9	1.6	boys 15.80 (± 1.42)	boys 16.96 (± 3.33)
					$15.93 (\pm 1.17)$	15.83 (± 3.33)
4	982	51.4	9.7	2.5	15.72 (± 1.66)	17.10 (± 3.79)
					15.70 (± 1.56)	15.39 (± 3.98)
5	953	51.1	12.7	2.7	15.62 (± 1.67)	17.11 (± 4.05)
					15.66 (± 1.61)	15.19 (± 4.46)
6	1006	51.4	13.5	3.5	15.88 (± 1.81)	17.26 (± 4.53)
					16.03 (± 2.08)	15.69 (± 5.36)
7	1026	51.4	17.1	5.1	16.24 (± 2.24)	18.24 (± 5.56)
					16.57 (± 2.59)	16.66 (± 6.28)
8	1037	51.2	20.7	6.3	16.85 (± 2.49)	19.46 (± 6.16)
					17.10 (± 2.86)	18.03 (± 7.35)
9	1067	51.4	20.1	5.2	17.64 (± 2.99)	21.46 (± 7.41)
					17.45 (± 2.74)	18.81 (± 7.85)
10	1018	51.4	22.6	4.9	18.29 (± 3.37)	22.61 (± 7.61)
					18.37 (± 3.17)	21.19 (± 9.26)

responders. The lower response proportion in resident aliens was compensated by oversampling.¹⁹ Of participating children, 8023 were between 3 and 10 years of age (Table 1). For 7767 of these, both SDP and BMI z-score were available, and for 7641, both SDP and KFA z-score were available, leaving between 844 and 1036 in each age. Figure 1 shows the amount of sleep decreasing with age.

Table 2 shows the associations of the primary explanatory variable SDP, the outcome measures, and selected potential confounders. Maternal BMI and birth weight were both inversely correlated with SDP and correlated with the outcome measures BMI and KFA z-score. In effect, if not adjusted, this can lead to an overestimation of the (inverse) association of sleep duration and body fat and mass. However, these correlations were weak. The KiGGS data confirm BMI and KFA increasing with decreasing SES. Children with a SES rated as middle showed the longest age specific sleep durations. High electronic media consumption was associated with lower sleep duration. Looking at the outcome measures, dose-response relationships to smoking in pregnancy and electronic media consumption are seen. The BMI and KFA z-scores correlated with r = 0.837 (P < 0.0001).

Results from the multivariable regression models for the BMI z-score as outcome are presented in Figure 2 (left part) and Table 3. Figure 2 illustrates, that the effects of SDP do not follow a systematic age dependency. Likewise, the interaction SDP*age is not significant (P = 0.873). It therefore seems justified to drop the interaction from the final model and to report a common effect estimate for all ages. The same applies to the model for the KFA z-score as outcome (see Fig. 2 [right part] and Table 4). For both outcomes, there was a significant association with SDP adjusted for confounders.

Interpretation of **B SDP**

Since the continuous variable SDP ranges from 0 to 1, its coefficient β compares the children with the highest to those with

the lowest sleep duration in an age group. To get a more intuitive measure, the following approximation is suggested: one hour of additional sleep which roughly corresponds to a upward shift of 30 points on the percentile scale is associated with a 0.30 $*\beta = 0.30 * (-0.235) = 0.071$ lower BMI z-score. For a 10-year-old girl, this would equal to a change in BMI of -0.237, given a 3.37 standard deviation for BMI in 10-year-old females.

For the sake of easier interpretation of the effect size of sleep on BMI and KFA z-score, we fit the same model using sleep duration in hours, as originally collected, instead of SDP. Since this raises the problem of confounding by age (which is circumvented by SDP) we had to add age to the model. Again, there was

no systematic age dependency seen when plotting the effects of sleep duration for each age (in the same manner as in Fig. 2), and the interaction sleep duration*age was not significant and was therefore dropped from the final model. The resulting β -coefficients for sleep duration were -0.064 [95 % confidence interval -0.088; -0.039] and -0.054 [-0.080; -0.029], indicating a decrease of 0.064 in BMI z-score, and 0.054 in KFA z-score per additional hour of sleep. The unadjusted β -coefficients were -0.054 [-0.074; -0.035] and -0.050 [-0.070; -0.030], respectively.

Does Longer Sleep Duration Shift the Entire or Only Parts of the BMI/KFA Distribution?

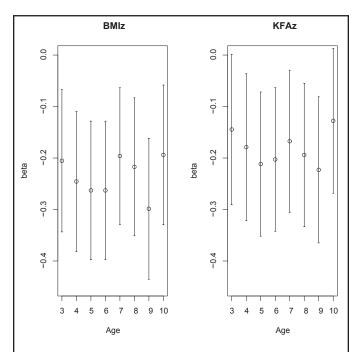
To assess whether the association of sleep duration and BMI is of the same sign and strength among the entire BMI distribution, we performed a quantile regression. Again taking the z-scores for all children from 3–10 y together, it seems that the middle and upper tail of the distribution are most affected by sleep duration (illustrated in Fig. 3, Table 5). The same is true for the distribution of KFA z-scores (Table 5). Table 5 lists quantile regression coefficients for the age-independent measure SDP as well as for sleeping hours as explanatory variables.

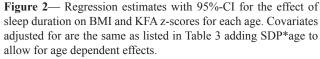
Since most studies on sleep duration and weight status report risk estimates, we fitted a logistic model, using the same covariates as presented earlier to provide odds ratios for being overweight (OR = 0.56 [0.42; 0.74]) or obese (OR = 0.51 [0.31; 0.84]). Analogous to the paragraph "*Interpretation of* β *SDP*" the OR for being overweight associated with an increase in sleep duration equivalent to 0.30 on the percentile scale can be computed as $0.56^{0.30} = 0.84$ [0.76; 0.91]. If the same model is fitted using sleep duration instead of SDP, the resulting OR for being overweight per additional hour of sleep is 0.84 [0.78; 0.91], which is in excellent agreement with the "rule of thumb" for the interpretation of β SDP given above. For obesity, this OR = 0.82 [0.70; 0.95].

Table 2-Potential Confounders of the Association Sleep Duration Percentile (SDP)-Overweight

Confounder	SDP		BMI z-score		KFA z-score	
maternal BMI	r, P -0.017, 0.124		r, P 0.253, < 0.0001		r, P 0.227, < 0.0001	
birth weight	-0.025, 0.027		0.175, < 0.0001		0.094, < 0.0001	
	mean ± SEM		mean ± SEM		mean ± SEM	
SES		P = 0.002		P < 0.0001		P < 0.0001
upper	0.499 ± 0.006		-0.185 ± 0.014		-0.187 ± 0.015	
middle	0.512 ± 0.005		-0.032 ± 0.013		-0.021 ± 0.013	
lower	0.486 ± 0.006		0.149 ± 0.016		0.140 ± 0.016	
Smoking in pregnancy		P = 0.080		P < 0.0001		P < 0.0001
no	0.499 ± 0.003		-0.049 ± 0.009		-0.041 ± 0.009	
rarely	0.515 ± 0.009		0.195 ± 0.025		0.160 ± 0.025	
regular	0.481 ± 0.015		0.304 ± 0.048		0.244 ± 0.043	
Excl. breastfed		P = 0.244		P < 0.0001		P < 0.0001
< 4 m	0.508 ± 0.005		0.048 ± 0.012		0.077 ± 0.012	
\geq 4 m	0.501 ± 0.004		-0.083 ± 0.012		-0.109 ± 0.012	
electronic media consump	otion	P < 0.0001		P < 0.0001		P < 0.0001
low	0.531 ± 0.005		-0.116 ± 0.013		-0.135 ± 0.013	
middle	0.513 ± 0.006		-0.038 ± 0.014		-0.024 ± 0.014	
high	0.457 ± 0.006		0.145 ± 0.016		0.155 ± 0.016	

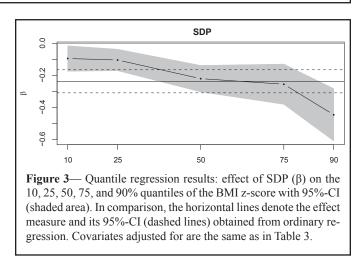
For metric variables the correlation coefficients, for categorical variables means for SDP, body mass index (BMI), and proportion body fat (KFA) z-score by level of the confounder are given. Symbols: correlation coefficient r, its P-value P, standard error of mean SEM.





DISCUSSION

Our data support the hypothesis that sleep duration is inversely associated with obesity in children between 3 and 10 years of age. To assess whether this effect is age specific, we constructed a measure SDP enabling us to compare sleep durations over age groups. When taking BMI or KFA z-score as out-



come, multivariable regression analysis revealed no systematic age dependency of the effect of sleep duration. Additionally, there are similar effects on BMI and KFA.

Interestingly, sleep duration is not homogeneously associated with BMI and KFA but mainly affects the upper tails of their distributions. This explains the different effect sizes obtained for sleep duration: the OR for overweight/obesity are pronounced while the changes in mean BMI are not impressive. It seems justified to publish OR as done in most publications on the subject from a biometric point of view. From the clinical perspective, the distinction of normal from not normal weight status also makes sense. However, if BMI itself is seen as a risk factor for e. g. cardiovascular diseases, it could be argued that even small shifts of the entire BMI distribution have great impact on the outcome on the population level. Therefore, modelling BMI as a metric outcome is interesting from a public health point of view. Furthermore, the fact, that sleep duration is associated with body composition primarily in children with higher

Table 3—Multivariable R	iable Regression Results for the BMI z-score				
Explanatory variable SDP	value/unit	β -0.235	standard error (β) 0.044	P-value < 0.0001	
social class	Upper middle (reference)	-0.042	0.029	0.001	
	Lower	0.074	0.028		
(maternal) BMI	in kg/m ²	0.045	0.003	< 0.0001	
smoking in pregnancy	no (reference)			< 0.0001	
	from time to time	0.220	0.036		
	Regularly	0.259	0.056		
exclusively breastfed at	no (reference)			0.105	
least until 4th month	Yes	-0.039	0.024		
Birthweight	in 500 g	0.130	0.010	< 0.0001	
electronic media	low (reference)			< 0.0001	
	intermediate	0.072	0.028		
	high	0.127	0.030		

For comparison, the unadjusted β for sleep duration percentile (SDP) is -0.273 [-0.356; -0.190].

Table 4---Multivariable Regression Results for the Proportion Body Fat (KFA) z-score

Explanatory variable SDP	value/unit	β -0.182	standard error (β) 0.046	P-value < 0.0001
social class	Upper	-0.075	0.030	< 0.0001
	middle (reference)			
	Lower	0.085	0.029	
(maternal) BMI	in kg/m ²	0.039	0.003	< 0.0001
smoking in pregnancy	no (reference)			< 0.0001
	from time to time	0.125	0.037	
	Regularly	0.214	0.058	
exclusively breastfed at	no (reference)			< 0.0001
least until 4th month	yes	-0.107	0.025	
birthweight	in 500 g	0.062	0.011	< 0.0001
electronic media	low (reference)			< 0.0001
	intermediate	0.104	0.029	
	high	0.152	0.031	

For comparison, the unadjusted β for sleep duration percentile (SDP) is -0.234 [-0.317; -0.150].

BMI and also is reflected by KFA helps to generate hypotheses or to trade off already proposed hypotheses on the underlying physiological mechanisms.

The similar findings in BMI and KFA support the hypothesis that it is actually fat (and not other tissues) that is mainly influenced by sleep. Also, children with higher BMI are likely to eat more and be less physically active. Therefore they might be more sensitive to hormonal imbalances caused by short sleep, because the metabolic dysregulation multiplies the opulence of food supply.

Not surprisingly, many previous hypotheses on potential mediators focus on hormonal changes induced by sleep duration: Growth hormone (GH) is secreted during sleep and promotes body height. Sleep deprived subjects show lower GH levels.²⁰ This could result in lower height and thus higher BMI. However, there was no positive association between SDP and height z-score (age and sex specific) in our data. Thus, the effect of sleep duration on weight status seems not to be mediated by length growth. Another effect of GH is the inhibition of adipose tissue lipoprotein lipase activity.21 It would be plausible, if subjects with higher body fat mass and thus higher amounts of this enzyme showed greater response to sleep induced GH changes. In summary, potential effects of decreased GH secretion related to sleep are more likely to be related to impact on lipoprotein lipase activity than length growth.

Ghrelin, an appetite stimulating hormone reported to be elevated by habitual short sleep duration,²⁰ is discussed as a potential mediator of the association of sleep duration and weight status. However, with ghrelin being secreted by the stomach and being lowered in obese subjects,²⁰ it can hardly explain the predominant effect of sleep duration on overweight subjects as found in the present study. While this does not rule out ghrelin being involved, our findings point to the relevance of other pathways regarding the association of sleep duration and weight status.

Leptin, an appetite lowering hormone, primarily secreted by adipose tissue is elevated in obesity and low-

ered in sleep deprivation. This mechanism is augmented by higher circulating levels of C-reactive protein (CRP), which binds leptin, found in obese and sleep-deprived subjects.^{22,23} Finally, this could mediate the stronger effect of sleep duration in subjects with higher body fat, as observed in our epidemiological data.

Regarding the last three pathways mentioned, a reanalysis of the data from previous laboratory studies stratified by weight status would be promising (in particular GH–lipoprotein lipase and leptin).

A strength of this study is the large sample size obtained from a survey nationally representative for Germany, a country with raising childhood obesity prevalence typical for the industrialized world.

The following limitations to our study have to be noted: sleep duration has been obtained from parental interview. The use of **Table 5**—Quantile Regression Results: Effects (β) of SDP (left part) and sleeping hours (right part) on different quantiles of BMI (upper part) and KFA (lower part) z-score

Quantile	SDP			sleeping hours			
	β	95%	6 CI	β	95%	6 CI	
BMI z-score							
10%	-0.093	-0.172	-0.013	-0.020	-0.041	0.002	
25%	-0.102	-0.169	-0.035	-0.029	-0.049	-0.009	
50%	-0.219	-0.301	-0.136	-0.055	-0.078	-0.031	
75%	-0.253	-0.379	-0.128	-0.072	-0.104	-0.040	
90%	-0.445	-0.608	-0.281	-0.128	-0.167	-0.090	
KFA z-score							
10%	-0.064	-0.131	0.003	-0.019	-0.036	-0.001	
25%	-0.062	-0.133	0.009	-0.011	-0.033	0.010	
50%	-0.128	-0.220	-0.036	-0.038	-0.065	-0.011	
75%	-0.240	-0.373	-0.107	-0.060	-0.097	-0.022	
90%	-0.222	-0.416	-0.027	-0.074	-0.122	-0.026	

Covariates adjusted for are the same as in Table 3. When using sleeping hours as explanatory variable, additional adjustment for age was done as explained under Interpretation of β SDP, 2nd paragraph.

actigraphy or 24-h recalls are alternatives that might result in a more valid measurement but are not feasible in such a large scale study. For the same reason we had to rely on a simple measure of body fat by skin fold thickness, which despite of some imprecision on the individual level correlates well with reference methods and is established in epidemiological studies.²⁴⁻²⁶

A comprehensive discussion on the causality of sleep duration and obesity is beyond the scope of this article and can be found in a review by Marshall and colleagues.²⁷ Since the present study is cross-sectional in design, reverse causation might be an issue. However, results of a recent longitudinal study in younger children support short sleep duration as a causal factor.⁵

Conclusions

Sleep needs change substantially within the first decade. The present study introduces a method to quantitatively compare the effect of sleep duration on weight status between different ages. Applied on data from a large, nationally representative sample it reveals no significant age dependency of the effect, which in agreement with a recent review²—is seen up to an age of 9 to 10 years. Sleep duration is associated with higher body fat mass resulting in higher BMI and mainly affects heavier children, who are in the focus of clinical interest. Further research to understand the association of sleep duration and obesity should consider the alignment of physiological hypotheses with these epidemiological findings, especially the interaction with weight.

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DISCLOSURE STATEMENT

This was not an industry supported study. Dr. Wabitsch has participated in speaking engagements for Pfizer. The other authors have indicated no financial conflicts of interest.

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