



High sugar content and body mass index: modelling pathways around the first 1000 d of life, BRISA cohort

Dâmaris Alves Silva Pinto¹, Joelma Ximenes Prado Teixeira Nascimento², Luana Lopes Padilha¹, Sueli Ismael Oliveira da Conceição², Ana Karina Teixeira da Cunha França², Vanda Maria Ferreira Simões¹, Rosângela Fernandes Lucena Batista¹, Marco Antônio Barbieri³ and Cecília Claudia Costa Ribeiro^{1,*} 

¹Programa de Pós-Graduação em Saúde Coletiva, Universidade Federal do Maranhão, 155 Barão de Itapary–Centro, São Luís, MA 65020-070, Brasil: ²Departamento de Ciências Fisiológicas, Curso de Nutrição, Universidade Federal do Maranhão, São Luís, MA, Brasil: ³Departamento de Puericultura e Pediatria, Faculdade de Medicina de Ribeirão Preto, Universidade de São Paulo, Ribeirão Preto, SP, Brasil

Submitted 15 April 2020: Final revision received 15 December 2020: Accepted 21 December 2020: First published online 28 December 2020

Abstract

Objective: Few studies are focused on sugar consumption around the first 1000 d of life. Thus, this work modelled the pathways linking the consumption of sugary drinks in pregnancy and maternal pre-gestational BMI to early child's exposure to products with high sugar content and to BMI z-score in the second year of life.

Design: BRISA cohort, São Luís, Brazil was used from the baseline to the follow-up at the second year of life.

Setting: A theoretical model was constructed to analyse associations between variables from prenatal period (socio-economic status, age, frequency of sugary drinks consumption during pregnancy and pre-gestational BMI), birth weight, exclusive breast-feeding and two outcomes: higher calories from products with added sugar as a percentage of the total daily energy intake and BMI z-score at follow-up at the first 2 years of life, using structural equation modelling.

Participants: Data of pregnant women (*n* 1136) and their offspring.

Results: Higher pre-gestational BMI (standardised coefficient (SC) = 0.100; *P* = 0.008) and higher frequency of sugary drinks consumption during pregnancy (SC = 0.134; *P* < 0.001) resulted in high percentage of daily calories from products with added sugar in the second year of child, although no yet effect was observed on offspring weight at that time.

Conclusions: Maternal obesity and sugary drinks consumption in pregnancy increased the risk of early exposure (before to 2 years) and high exposure of child to added sugar, showing perpetuation of the unhealthy dietary behaviours in the first 1000 d of life.

Keywords

Added sugar
Gestation

1000 d of life

Structural equation modelling

Obesity represents a global public health problem^(1,2) and an increasing proportion of women start pregnancy with excess weight or obesity⁽³⁾. Maternal obesity has an important role in the development of child obesity⁽³⁾, being observed in the first 1000 d of life – period from conception until 2 years of age⁽⁴⁾. Simultaneously, child obesity has reached epidemic levels in developed and developing countries^(5,6), taking into account that most of the excess weight in childhood is gained in the preschool years⁽⁷⁾.

In view of obesity prevention, the WHO published a guideline suggesting reduction in sugar intake, recommending that it does not exceed 10%, ideally 5%, of the consumption of total daily calories⁽⁸⁾. In order to prevent future cardiovascular risk, the American Heart Association also published a guideline suggesting the non-exposure to added sugar in the first 2 years of life⁽⁹⁾.

Sugary drinks, such as sodas and fruit juices, are the main source of discretionary calories in diet (the excess

*Corresponding author: Email cecilia_ribeiro@hotmail.com

© The Author(s), 2021. Published by Cambridge University Press on behalf of The Nutrition Society

calories ingested after to meet recommended nutrient intakes)⁽¹⁰⁾, which consumption has been consistently associated with overweight, even in children^(11–14). Although the early exposure to sugars is a cause of serious concern, few studies have been published focusing on sugar-sweetened beverages consumption in the first 1000 d of life^(15–17). The first 1000 d includes gestation and the child's first 2 years, period that there is more phenotypic plasticity, when environmental influences may acts increasing the risk of obesity and non-communicable diseases in future⁽¹⁸⁾. The sugar-sweetened beverages consumption during pregnancy was associated with greater adiposity in preschool⁽¹⁵⁾ and middle school children⁽¹⁶⁾. Furthermore, the higher sugar-sweetened beverages intake during gestational period has shown linked to harmful effect in offspring health, such as risk of atopy and asthma^(19–21), while the children intake of sugar-sweetened beverages in the first year of life was associated with higher adiposity levels at 6 years of age⁽¹⁷⁾ and to asthma traits in children⁽²²⁾.

The mechanisms connecting the sugary drinks intake during the pregnancy and early exposures to offspring for products with high sugar content may be linked to social vulnerability^(23,24), paternal attitudes and unhealthy dietary behaviours⁽²⁵⁾. As physiological mechanisms, the higher gestational exposure to sugary drinks may result in epigenetic changes and fetal programming with effects along life course^(26,27). Furthermore, the hyperinsulinaemia in gestational period may result in childhood obesity^(28,29).

Structural equation modelling allows to test a complex hypothetical causal structure⁽³⁰⁾. Exposures from the perinatal period, such as sugar consumption and maternal obesity, may increase future risk to early exposure to sugar in offspring and higher obesity risk in their children; thus, modelling direct and indirect pathways linking those perinatal factors may add knowledge that will allow to better analyse this multicausal structure around the first 1000 d of life.

In this context, the current study modelled the pathways linking sugary drinks consumption during pregnancy and maternal pre-gestational BMI with the child's exposure to high sugar content products, as well as the child's growth in the second year of life.

Method

Here, the data from the prospective cohort of pregnant women BRISA in the city of São Luís, Brazil were used. The pregnant women were invited to participate in the study in prenatal services and in public and private hospitals from February 2010 to June 2011⁽³¹⁾.

At baseline, pregnant women who were held prenatal care from 22nd and 25th gestational week were included ($n = 1447$). In the occasion of birth, from May 2010 to November 2011, puerperal women ($n = 1382$) were interviewed again during the first 24 h after delivery (first

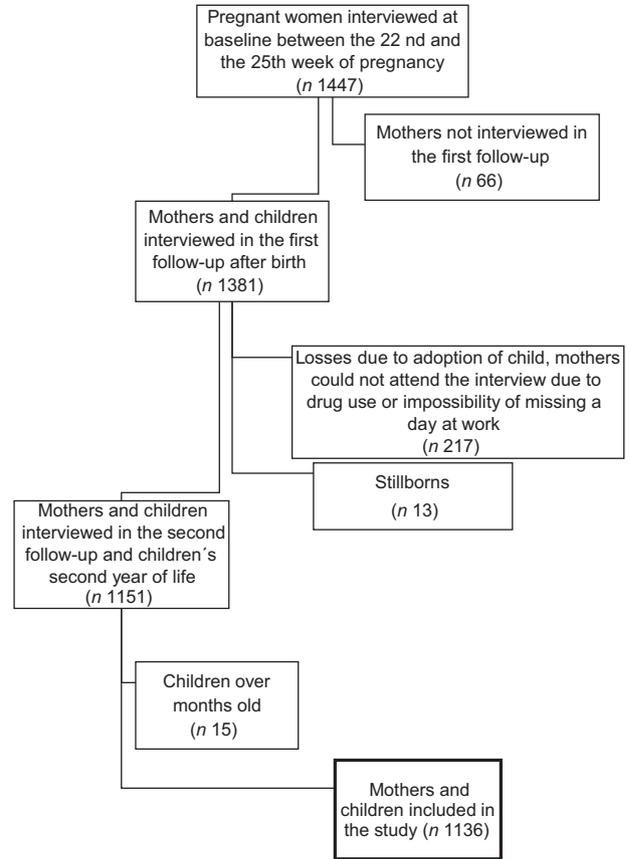


Fig. 1 Flow diagram of the BRISA cohort, São Luís, Brazil

follow-up). Mother and children ($n = 1151$) were evaluated between September 2011 and March 2013 (second follow-up). The study sample was composed of 1136 mother–child pairs, with the children up to 24 months of age (Fig. 1).

The following information was used from pregnant women at baseline: age (years), monthly family income (multiples of minimum wage), education (years of study), occupation of the head of the household, economic class according to the Brazilian Criteria of Economic Classification (A/B, C, D/E), ranging from E (lowest economic class) to A (highest economic class)⁽³²⁾, self-reported pre-gestational weight (kg), height (m) measured with a portable stadiometer (Altuxata[®])^(33,34); subsequently, the pre-gestational BMI⁽³⁴⁾ and frequency of consumption of soft drinks and chocolate beverages during current pregnancy were calculated.

The frequency of consumption of sugary drinks (soft drinks and chocolate beverages), from the 22nd to 25th gestational weeks, was calculated using the following questions from the prenatal questionnaire: ‘How many days per week do you drink soft drinks?’, ‘How many times per day do you drink soft drinks?’, ‘How many days per week do you drink chocolate beverages?’, ‘How many times per day do you drink chocolate beverages?’ and the weekly consumption (zero to seven times per week)

multiplied by the daily consumption (one to six times per day). Finally, the consumption frequencies for each sugary drink were added and this variable was analysed in categories: none, until 1 time per day, until to 2 times per day and 3 or more times per day.

Previous study using São Luis BRISA cohort data showed an association between this higher frequency of soft drink consumption in pregnancy with maternal periodontal diseases⁽³⁵⁾ and with asthma in offspring⁽¹⁹⁾.

A second follow-up of the cohort was conducted during the second year of life of the children (13–24 months old). From the second follow-up, the following information was used about the children: age (months), sex, weight (kg) measured with digital scale (Tanita®)⁽²⁹⁾, length/height (cm) measured with a portable stadiometer (Altuxata®)⁽²⁹⁾, exclusive breast-feeding treated as a dichotomous categorical variable: <6 months or ≥6 months, the BMI in z-score⁽³⁶⁾, a continuous numerical variable and data from the food consumption through the application of 24-h food recalls reported for mother.

The nutritional composition of the child's diet in the second cohort follow-up was calculated with the *Software Virtual Nutri Plus*® (2010 version). Details from the data collection of food consumption in the second year follow-up of BRISA cohort are published in Padilha *et al.*⁽³⁷⁾. The percentage of calories from most frequently used industrialised products with high sugar content by

children was derived from non-dairy beverages (soft drinks, industrialised juices, nectars and reconstituted powdered juices); dairy sugary products (ready-to-drink chocolate beverages, ice creams, popsicles, creamy beverages, yogurts and foods of the Petit Suisse type) and industrialised sweetened puree baby food. Thus, the calories from products with added sugar as a percentage of the total daily energy intake were analysed in theoretical model by structural equation modelling. Subsequently, solid foods (cakes and cookies) were also tested replacing those products with a high sugar content and were analysed by structural equation modelling (data not shown in the tables).

Proposed theoretical model

The proposed theoretical model was constructed to analyse the association between frequency of sugary drinks consumption in pregnancy and maternal pre-gestational BMI to the outcomes of early child's exposure to products with a high sugar content and BMI z-score in the second year of life (Fig. 2). Socio-economic status (SES) was assumed as the more distal determinant (exogenous variable), exerting its effects on all other variables in model. Maternal age may influence maternal weight, the choice and supply of food for herself and her child, and the child BMI. Higher maternal exposure to sugary drinks may be related to higher maternal BMI and may link the pathways

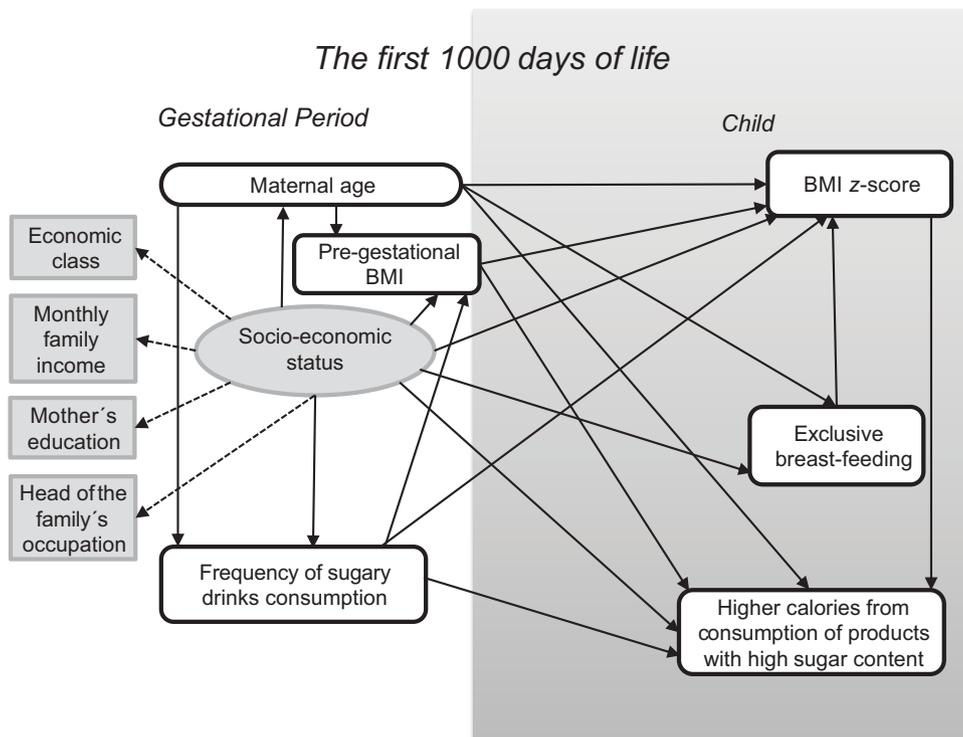


Fig. 2 Proposed theoretical model to evaluate the association between the consumption of sugary drinks in pregnancy and maternal pre-gestational BMI to early child's exposure to products with high sugar content and to BMI z-score in the second year of life in the BRISA cohort. São Luís-MA, 2010. ○, Latent variables; ▭, effect indicators of the latent variable; □, observed variables

to child's exposure to products with a high sugar content and higher child's BMI in the second year of life. Higher maternal BMI may be associated with early child's exposure to products with a high sugar content and to higher child's BMI in the second year of life. The higher maternal BMI may be associated with a shorter duration of exclusive breast-feeding, while exclusive breast-feeding may have a protective effect on the child's BMI.

Statistical methods

Structural equation modelling is an epidemiological tool that allows to test direct and indirect paths between multiple variables, observed, and also latent variables. Latent variables are derived from their effect indicator variables, representing the shared variance among them, estimating effects free from the measurement error bias⁽³⁰⁾.

The SES was latent variable family deduced through the following indicator variables collected at prenatal period: a) mother's education, b) occupation of the head of the family, c) monthly household income relative to the Brazilian national minimum wage in 2010 and d) economic class. This variable has already been analysed in a previous study by BRISA prenatal cohort⁽³⁸⁾.

The initial statistical analyses and construction of the weighting variable were performed using the Stata[®] 14.0 software, and the Mplus[®] 7.0 software was used for structural equation modelling.

For the good model adjustment, the following estimates were considered: a) $P > 0.05$ and an upper limit of the CI of $90\% < 0.08$ for the root mean square error of approximation, b) comparative fit index and Tucker–Lewis index > 0.95 and c) value < 1 for the weighted least squares mean and variance adjusted. If these parameters were not adequate to the originally proposed model adjustment, the modindices command would be used in the structural equation modelling⁽³⁰⁾ to identify new pathways which if included may result in a better fitting model. The command modindices was used to calculate the modification indices (higher than 10 values) and thus to identify new paths that could improve the model fit⁽³⁹⁾. The total (the sum of direct and indirect effects), direct (not mediated) and indirect (through the mediator) effects were estimated, adopting effect criterion if $P < 0.05$ ⁽³⁰⁾.

In order to minimise the possibility of selection bias from the losses of the sample and to increase the external validity of the study, the analysis was weighted by the inverse probability of participation, comparing data of the mothers who participated in the follow-up with those of the mothers who did not participate through the χ^2 test. Fewer children who were born at an earlier gestational age attended the second follow-up visit. On this basis, the sample was weighted by calculating the probability of a children being seen at the second follow-up visit as a function of gestational age using a logistic regression model. The inverse of this probability of selection was then calculated and this variable was used to weight the SEM estimates.

Results

The average of the pre-gestational BMI was 23.1 kg/m² (SD 4.1) and 23.6% of the pregnant women presented excess weight. The frequency of sugary drinks consumption during pregnancy 58.0% had sugary drinks intake at least once a day, 18.7% until to 2 times per day and 8.8% three or more times a day, the average age of the mothers was 26.1 (SD 5.6) years and 77.2% of the mothers studied for 9–11 years. In 40.5% of the cases, the head of the family had a manual semi-specialised occupation, 46.0% of the mothers belonged to families with average monthly income of 1 to < 3 minimum wages and approximately 65.6% of them belonged to class economic C (Table 1).

Table 1 Socio-demographic, economic and nutritional characteristics of pregnant women in the BRISA prenatal cohort, São Luís – MA, 2010–2013 (n 1136)

Variables	<i>n</i>	%
Mother's age (years)		
Mean	26.1	
SD	5.6	
Median	26.0	
Percentiles 25–75	22.0–30.0	
Pre-gestational BMI (kg/m ²)		
Mean	23.1	
SD	4.1	
Median	22.4	
Percentiles 25–75	20.3–25.2	
Mother's education (years)		
0–4	17	1.5
5–8	114	10.0
9–11	877	77.2
≥ 12	127	11.1
No information*	1	0.2
Occupation of the head of the family		
Manual non-qualified	308	27.1
Manual semi-specialised	460	40.5
Manual specialised	51	4.5
Office positions	161	14.2
Higher education professional	58	5.1
Administrators/managers/directors/proprietaries	37	3.3
No information*	61	5.3
Family income (minimum wages)†		
< 1	12	1.0
1 and < 3	523	46.0
3 and < 5	362	31.9
≥ 5	208	18.3
No information*	31	2.8
Economic class‡		
A–B (upper class)	169	14.9
C (middle class)	745	65.6
D–E (lower class)	170	14.9
No information*	52	4.6
Sugary drinks		
None	164	14.4
Until 1 time per day	659	58.0
Until to 2 times per day	213	18.7
3 or more times per day	100	8.8

*Values unknown or not informed.

†Monthly family income based on Brazilian national minimum wage (approximately US\$ 290.00 in 2010).

‡Economic class according to the Brazilian Criteria of Economic Classification, categorised in A/B, C and D/E (BRASIL, 2008).

**Table 2** Characteristics of children at second follow-up in the BRISA prenatal cohort, Sao Luis – MA, 2010–2013 (*n* 1136)

Variables	<i>n</i>	%
Child's age (months)		
Mean	15.9	
SD	2.1	
Median	15.0	
Percentiles 25–75	14.0–17.0	
BMI†		
Mean	0.6	
SD	1.3	
Median	0.6	
Percentiles 25–75	–0.2–1.4	
Exclusive breast-feeding (months)		
< 6	501	44.1
≥ 6	614	54.0
No information*	21	1.9

*Values unknown or not informed.

†z-score of BMI for age (reference: WHO, 2006).

Among the children, 54.0% were exclusively breastfed for 6 months, the average of the z-score of BMI for age was 0.6 kg/m² (SD 1.3) (Table 2) and more than half of the children (54.8%) with an average age of 15.9 (SD 2.1) months was already exposed to products with a high sugar content. Among them, 37.0% consumed more than 5.0% of total calories from the sugar that is present in those products in the first 1000 d of life (Table 3).

The latent variable SES showed a good fit, with all of the indicators with factorial loads above 0.50 and with significant *P*-value (*P* < 0.001). Similarly, the final model analysed in structural equation modelling showed a good fit for all of the indexes considered (root mean square error of approximation: 0.023, 90% CI 0.000, 0.040, *P*-value root mean square error of approximation = 0.998, comparative fit index: 0.988, Tucker–Lewis index: 0.976 and weighted root mean square residual: 0.597) (Table 4). Suggestions of modification for this model were not theoretically plausible; thus, the analysis of the model proceeded as originally proposed.

Higher values for the SES explained the higher maternal age at pregnancy (standardised coefficient (SC) = 0.111; *P* = 0.012) and indirect association of protection of the SES for the higher frequency of consumption of sugary drinks during pregnancy via the mother's older age (SC = –0.015; *P* = 0.037) (Table 5).

The mother's older age was associated with an increase in pre-gestational BMI (SC = 0.260; *P* < 0.001), although older age in pregnancy has a direct association of protection for excess weight in the second year (SC = –0.090; *P* = 0.021) (Table 5). An indirect association was also observed for the increase in z-score of BMI in the child through a higher pre-gestational BMI (SC = 0.030; *P* = 0.008) (Table 5).

The higher pre-gestational BMI (SC = 0.100; *P* = 0.008) and the higher frequency of sugary drinks consumption during pregnancy (SC = 0.134; *P* < 0.001) resulted in higher percentage of intake of calories from products with high

sugar content at the second year of life, revealing an obesogenic environment in the first 1000 d of life (Table 5).

The higher maternal BMI was not associated with a shorter duration of exclusive breast-feeding (SC = –0.040; *P* = 0.438), nor exclusive breast-feeding had protective effect on the child's BMI (SC = –0.080; *P* = 0.095) (data not shown in the table). The higher percentage of calories from products with high sugar content by the children did not result in excess weight in the second year of life.

Our findings showed that the higher pre-gestational BMI were associated to the higher z-score BMI in the offspring (SC = 0.103; *P* = 0.006) (Table 5).

Mother's sugary drinks consumption was also associated with higher exposure to solid foods containing sugar (SC = 0.132; *P* < 0.001). However, maternal obesity was not associated with higher exposure to solid foods containing sugar (SC = 0.003; *P* = 0.715) (data not shown in the tables).

Discussion

In the present study, the higher pre-gestational BMI and the higher frequency of sugary drinks consumption during pregnancy resulted in higher percentage of intake of calories from added sugar in the second year of life, revealing behavioural and metabolic risk factors in the first 1000 d of life.

More than half of the children (54.8%) with an average age of 15.9 (SD 2.1) months was already exposed to industrialised products with a high sugar content, while according to recommendations from the American Heart Association children under 2 years should avoid consuming any added sugar⁽⁹⁾. Furthermore, the average consumption of calories from sugar in relation to total calories was 7.4%, which is considered high, since it is above the 5% recommended as ideal for the prevention of obesity according to the WHO⁽⁸⁾. In addition, 37.0% of the sample exceeded the 5% of the total calories coming only from products with a high sugar content, with even higher risk of obesity in the future⁽⁸⁾.

The findings that the higher pre-gestational BMI explains the higher consumption of calories from added sugar by the offspring reveal the role of environmental factors in obesity that are already present in the first 1000 d of life. These data support the findings of a previous study, showing that children of obese mother have an increased risk for obesogenic eating practices in the beginning of life⁽³⁾.

The higher frequency of sugary drinks consumption during gestation resulted in higher percentage of intake of calories from products with high sugar content by the offspring (SC = 0.134; *P* < 0.001), reinforcing again the role of behavioural and environmental factors in obesity risk. The children's eating preferences are shaped by a combination of genetic and environmental factors or both^(40–42), showing that mothers perform an important role in the eating

Table 3 Consumption of products with a high sugar content data of the children. Mean, SD, median and percentiles. BRISA prenatal cohort, São Luís, Brazil, 2010–2013

Variable	Mean	SD	Median	Percentiles 25–75
Consumption of products with a high sugar content (%)*:	7.4	5.6	5.7	3.3–9.8
Chocolate milk (kJ)†	361.1	237.2	334.7	167.8–502.1
Soft drinks (kJ)†	238.1	114.2	217.6	167.4–296.2
Industrialised juices (kJ)†	306.7	203.8	263.3	179.9–343.9
Sugared pasty products (%)*:	7.3	4.6	6.3	4.1–8.7
Sugary yogurts (kJ)†	325.5	194.5	330.5	165.3–330.5
Popsicles and ice cream (kJ)†	340.6	129.3	383.2	184.1–406.7
Industrialised baby food (kJ)†	297.9	58.2	318.0	268.8–319.2
Percentage of sugar intake in products with high sugar in relation to total calorie				
		<i>n</i>	<i>%</i>	
None		330	45.2	
< 5 %		130	17.8	
5 to < 10 %		175	23.9	
≥ 10 %		96	13.1	
Total		731	100.0	

*Percentage of daily calories from added sugar.

†Total energy of sugars.

Table 4 Indexes of expected and found model adjustments. BRISA cohort, São Luís – MA, 2010–2013

Indexes of model adjustment	Expected index values*	Index values found in the model†
χ^2 test		33.414
DF		23
<i>P</i> -value χ^2 test		0.0741
Root mean square error of approximation	< 0.05	0.023
90 % CI	< 0.08	0.000–0.040
<i>P</i> -value root mean square error of approximation	> 0.05	0.998
Comparative fit index	> 0.95	0.988
Tucker–Lewis index	> 0.95	0.976
Weighted Root Mean Square Residual	< 1.0	0.597

*Expected index values for a good fit of the model.

†Final model analysed by structural equation modelling.

practices of their children^(40,41). Maternal consumption of sugary drinks after pregnancy could be an indicator of the child's early exposure to products with high sugar content, however, unfortunately these data were not assessed in BRISA Cohort.

The higher percentage of sugar consumption through added sugar by the children did not result in excess weight in the second year of life. This finding is different from the expected, since previous studies showed association of early exposures to sugar-sweetened beverages with excess weight in childhood^(43,44). Children exposed to sugar-sweetened beverages at 2 years of age had prospectively an increase in BMI in the two following years⁽⁴³⁾. Early exposure to sugar-sweetened beverages increased the risk of obesity at 6 years of age, being a risk 92 % higher than the non-exposure when the introduction of beverages occurred before 6 months of age⁽¹⁷⁾. Therefore, it is possible that the effect of added sugar

consumption on excess weight is observed later in life, since systematic reviews show consistent association between sugar-sweetened beverages consumption and excess weight, however, in children who are older than in those in the present study^(11–14).

The average proportion of calories derived from added sugars consumed by the children was higher than 5 % that has been recommended by the WHO for the prevention of obesity.⁽⁸⁾ The American Heart Association also recommends to avoid consumption of added sugars by children up to 2 years to reduce cardiovascular risk in future⁽⁹⁾.

Our findings that the higher pre-gestational BMI was associated with the higher BMI *z*-score in the offspring validate the previous data that showed pre-gestational obesity as a risk factor for child obesity in the future^(44,45). The influence of pre-gestational excess weight in the development of childhood obesity can reveal itself through an interaction of genetic, epigenetic and environmental factors^(46–48). There is evidence of a strong genetic influence in the appetite of children, but the environment also has an important role in shaping the eating behaviours in childhood⁽⁴⁰⁾. This fact could be especially relevant when taking into account the taste of sugar^(49,50), which can result in compulsive eating behaviours⁽⁴⁶⁾.

Higher values for the SES were associated here with the higher consumption of sugary drinks during pregnancy. The data contrast with previous studies, in which lower SES was associated with higher consumption of soft drinks in European residences⁽⁵¹⁾ and among Hispanic pregnant women from low-income households in the USA⁽¹⁵⁾. The different findings from our study could be explained by the increase in the soft drinks consumption that is still occurring in populations in low- or medium-income countries⁽⁵²⁾. In Brazil, the consumption of these beverages is still increasing⁽⁵³⁾, especially among higher income groups⁽⁵⁴⁾.



Table 5 Standardised coefficient, SE and P-value of total and direct effects for the indicator variables. São Luís – MA, 2010–2013

Explanatory variables	Outcome	Total effects			Direct effects			Main indirect effects		
		SC	SE	P	SC	SE	P	SC	SE	P
SES*	Mother's age	0.111	0.044	0.012	0.111	0.044	0.012	0.000	0.000	1.000
SES*	Mother's sugary drinks consumption	0.037	0.045	0.409	0.052	0.045	0.251	-0.015	0.007	0.037¶
Mother's age	Pre-gestational BMI‡	0.260	0.015	< 0.001	0.255	0.016	< 0.001	0.006	0.006	0.354
Mother's age	Child's BMI z-score§	-0.061	0.038	0.113	-0.090	0.039	0.021	0.030	0.011	0.008**
Pre-gestational BMI‡	Child's BMI z-score§	0.103	0.038	0.006	0.106	0.039	0.006	-0.002	0.006	0.724
Pre-gestational BMI‡	Children's added sugar consumption	0.100	0.038	0.008	0.100	0.038	0.008	0.000	0.000	1.000
Mother's sugary drinks consumption	Children's added sugar consumption	0.134	0.028	< 0.001	0.138	0.028	< 0.001	-0.004	0.005	0.381

SC, standardised coefficient.

*Latent variable of prenatal socio-economic status.

†Frequency of consumption sugary drinks in pregnancy (soft drinks and chocolate beverages).

‡Pre-gestational BMI.

§z-score of BMI for age.

||Percentage of daily calories from added sugar (products with a high sugar content).

¶Indirect association of protection of the SES for the higher frequency of consumption of sugary drinks during pregnancy via the mother's older age.

**Indirect association was also observed for the increase in z-score of BMI in the child through a higher pre-gestational BMI.

The increased added sugar consumption by industrialised products has resulted in the proposal of regulatory measures, such as taxation, warning messages in labels of products with high sugar content and public policies to fight obesity⁽⁵⁵⁾. The data in the present study are even more concerning, since the children's consumption in the second year of life was high according to the WHO guidelines for obesity prevention⁽⁸⁾. Regulatory policies for sugary drinks could also benefit pregnant women, considering that the intake of these beverages during pregnancy is also concerning, since it could result in pre-eclampsia^(56,57) and preterm birth^(58,59).

As limitations of the study, it is important to emphasise the use of convenience sampling in the prenatal due to the impossibility of obtaining a random sample that is representative of the population of pregnant women in São Luís, Brazil. The representativity of the sample of this BRISA prenatal cohort was already compared with the populational base sample of the BRISA birth cohort performed during the same period in the city of São Luís; merely, the intermediate categories of mother's education (5–8 years and 9–11 years of study) were greater in the present sample, while the frequencies of the other variables were similar, reinforcing the external validity of our data⁽¹⁹⁾. Yet, in this research only the frequency of sugary drinks consumption during the pregnancy was available in the questionnaire applied in gestational period. Furthermore, the consumption variables had missing data; however, Mplus software imputes values for missing based on the variables that preceded it in the path analysis, using frequency analysis and Bayesian analysis⁽⁶⁰⁾.

As far as the strengths of the study, we highlight the use of data from a prospective cohort of pregnant women, which allowed for the observation of the sample in two moments: during pregnancy and the child's second year of life. Also, the use of structural equation modelling allowed the analysis of total, direct and indirect pathways between the exposure to products with high sugar content and excess weight around the first 1000 d of life.

Finally, the higher pre-gestational BMI and higher frequency of sugary drinks consumption during gestation were identified in the current study as the pathways that result in early (before to 2 years) and high exposure to added sugar in offspring, reinforcing the importance of the recommendation to avoid the consumption of these products in the first 1000 d of life. Strategies to promote adequate and healthy eating habits must be directed to pregnant women and to the families of those children, sensibilising them to the adverse effects of added sugar consumption on people's health in a transgenerational way.

Acknowledgements

Acknowledgements: Not applicable. *Financial support:* Maranhão State Research and Scientific and Technological Development Foundation (FAPEMA), São Paulo State

Research Foundation (FAPESP), and National Council for Scientific and Technological Development (CNPq) and the Coordination for the Improvement of Higher Education Personnel (CAPES), 'Finance Code 001'. *Conflict of interest:* No competing financial interests exist. *Authorship:* Contributions of the authors: D.A.S.P., J.X.P.T.N., L.L.P., S.I.O.C., A.K.T.C.F., V.M.F.S., R.F.L.B., M.A.B. and C.C.C.R. Analysed and interpreted data: D.A.S.P., J.X.P.T.N., L.L.P. and C.C.C.R. Elaborated the manuscript: D.A.S.P., J.X.P.T.N., L.L.P., S.I.O.C., A.K.T.C.F., V.M.F.S., R.F.L.B., M.A.B. and C.C.C.R. Critical reading of the manuscript and final approval of the version to be published: D.A.S.P., J.X.P.T.N., L.L.P., S.I.O.C., A.K.T.C.F., V.M.F.S., R.F.L.B., M.A.B. and C.C.C.R. *Ethics of human subject participation:* 'This study was conducted according to the guidelines laid down in the Declaration of Helsinki and all procedures involving research study participants were approved by the Ethics Committee from the Hospital of Federal University of Maranhão'. Written informed consent was obtained from all subjects/patients.

References

- World Health Organization (WHO) (2013) *Obesity and overweight, Fact sheet N 311, updated March 2013*. Geneva: WHO. <https://www.who.int/news-room/fact-sheets/detail/obesity-and-overweight> (accessed February 2018).
- Ng M, Fleming T, Robinson M *et al.* (2014) Global, regional, and national prevalence of overweight and obesity in children and adults during 1980–2013: a systematic analysis for the Global Burden of Disease Study 2013. *Lancet* **384**, 766–781.
- Thompson AL (2013) Intergenerational impact of maternal obesity and postnatal feeding practices on pediatric obesity. *Nutr Rev* **71**, S55–S61.
- Baidal JAW, Locks LM, Cheng ER *et al.* (2016) Risk factors for childhood obesity in the first 1000 days: a systematic review. *Am J Prev Med* **50**, 761–779.
- Sahoo K, Sahoo B, Choudhury AK *et al.* (2015) Childhood obesity: causes and consequences. *J Family Med Prim Care* **4**, 187–192.
- Abarca-Gómez L, Abdeen ZA, Hamid ZA *et al.* (2017) Worldwide trends in body-mass index, underweight, overweight, and obesity from 1975 to 2016: a pooled analysis of 2416 population-based measurement studies in 128·9 million children, adolescents, and adults. *Lancet* **390**, 2627–2642.
- Lanigan J (2018) Prevention of overweight and obesity in early life. *Proc Nutr Soc* **77**, 247–256.
- World Health Organization (WHO) (2015) Guideline: Sugars Intake for Adults and Children. World Health Organization. https://www.who.int/nutrition/publications/guidelines/sugars_intake/en/ (accessed February 2018).
- Vos MB, Kaar JL, Welsh JA *et al.* (2017) Added sugars and cardiovascular disease risk in children: a scientific statement from the American Heart Association. *Circulation* **135**, e1017–e1034.
- Johnson RK, Appel LJ, Brands M *et al.* (2009) Dietary sugars intake and cardiovascular health: a scientific statement from the American Heart Association. *Circulation* **120**, 1011–1020.
- Malik VS, Pan A, Willett WC *et al.* (2013). Sugar-sweetened beverages and weight gain in children and adults: a systematic review and meta-analysis. *Am J Clin Nutr* **98**, 1084–1102.
- Keller A & Bucher Della Torre S (2015) Sugar-sweetened beverages and obesity among children and adolescents: a review of systematic literature reviews. *Child Obes* **11**, 338–346.
- Della Torre SB, Keller A, Depeyre JL *et al.* (2016) Sugar-sweetened beverages and obesity risk in children and adolescents: a systematic analysis on how methodological quality may influence conclusions. *J Acad Nutr Diet* **116**, 638–659.
- Luger M, Lafontan M, Bes-Rastrollo M *et al.* (2017) Sugar-sweetened beverages and weight gain in children and adults: a systematic review from 2013 to 2015 and a comparison with previous studies. *Obes Facts* **10**, 674–693.
- Watt TT, Appel L, Roberts K *et al.* (2013) Sugar, stress, and the Supplemental Nutrition Assistance Program: early childhood obesity risks among a clinic-based sample of low-income Hispanics. *J Commun Health* **38**, 513–520.
- Gillman MW, Rifas-Shiman SL, Fernandez-Barres S *et al.* (2017) Beverage intake during pregnancy and childhood adiposity. *Pediatrics* **140**, e20170031.
- Pan L, Li R, Park S *et al.* (2014) A longitudinal analysis of sugar-sweetened beverage intake in infancy and obesity at 6 years. *Pediatrics* **134**, S29–S35.
- Barker DJ, Eriksson JG, Forsen T *et al.* (2002) Fetal origins of adult disease: strength of effects and biological basis. *Int J Epidemiol* **31**, 1235–1239.
- Nascimento JXPT, Ribeiro CCC, Batista RFL *et al.* (2017) The first 1000 days of life Factors associated with "Childhood Asthma Symptoms": Brisa Cohort, Brazil. *Sci Rep* **7**, 16028.
- Bédard A, Northstone K, Henderson AJ *et al.* (2017) Maternal intake of sugar during pregnancy and childhood respiratory and atopic outcomes. *Eur Respir J* **50**, 1700073.
- Wright KM, Dono J, Brownbill AL *et al.* (2019) Sugar-sweetened beverage (SSB) consumption, correlates and interventions among Australian Aboriginal and Torres Strait Islander communities: a scoping review. *BMJ Open* **9**, e023630.
- Padilha LL, Vianna EO, Vale ATM *et al.* (2020) Pathways in the association between sugar sweetened beverages and child asthma traits in the 2nd year of life: findings from the BRISA cohort. *Pediatr Allergy Immunol*. Published online: 11 March 2020. doi: 10.1111/pai.13243.
- De Coen V, Vansteelandt S, Maes L *et al.* (2012) Parental socioeconomic status and soft drink consumption of the child. The mediating proportion of parenting practices. *Appetite* **59**, 76–80.
- Pettigrew S, Jongenelis M, Chapman K *et al.* (2015) Factors influencing the frequency of children's consumption of soft drinks. *Appetite* **91**, 393–398.
- Woo Baidal JA, Morel K, Nichols K *et al.* (2018) Sugar-sweetened beverage attitudes and consumption during the first 1000 days of life. *Am J Public Health* **108**, 1659–1665.
- Mateo-Fernández M, Merinas-Amo T, Moreno-Millán M *et al.* (2016). In vivo and in vitro genotoxic and epigenetic effects of two types of cola beverages and caffeine: a multiassay approach. *Biomed Res Int* **2016**, 7574843.
- de Boo HA & Harding JE (2006) The developmental origins of adult disease (Barker) hypothesis. *Aust N Z J Obstet Gynaecol* **46**, 4–14.
- Reichetzedder C, Dwi Putra SE, Li J *et al.* (2016) Developmental origins of disease – crisis precipitates change. *Cell Physiol Biochem* **39**, 919–938.
- Jen V, Erler NS, Tielemans MJ *et al.* (2017) Mothers' intake of sugar-containing beverages during pregnancy and body composition of their children during childhood: the Generation R Study. *Am J Clin Nutr* **105**, 834–841.



30. Kline RB (2016) *Principles and Practice of Structural Equation Modeling*, 3rd ed. New York, NY: The Guilford Press.
31. da Silva AAM, Simões VMF, Barbieri MA *et al.* (2014) A protocol to identify non-classical risk factors for preterm births: the Brazilian Ribeirão Preto and São Luís prenatal cohort (BRISA). *Reprod Health* **11**, 79.
32. Research Companies Brazilian Association (2012) Brazilian Economic Classification Criteria. São Paulo: ABEP. <http://www.abep.org/criterio-brasil> (accessed February 2018).
33. Lohman TG, Roche AF & Martorell R (1988) *Anthropometric Standardization Reference Manual*, Vol. 177. Champaign, IL: Human kinetics books Champaign.
34. World Health Organization (WHO) (1995) Physical Status: the Use and Interpretation of Anthropometry. Geneva: WHO Technical Report Series no. 854. https://www.who.int/childgrowth/publications/physical_status/en/ (accessed February 2018).
35. Menezes CC, Ribeiro CCC, Alves CMC *et al.* (2019) Soft drink consumption and periodontal status in pregnant women. *J Periodontol* **90**, 159–166.
36. World Health Organization (WHO) (2006) WHO child growth standards: length/height for age, weight-for-age, weight-for-length, weight-for-height and body mass index-for-age, methods and development. https://www.who.int/childgrowth/standards/technical_report/en/ (accessed February 2018).
37. Padilha LL, França AKTC, da Conceição SIO *et al.* (2017) Nutrient intake variability and the number of days needed to estimate usual intake in children aged 13–32 months. *Br J Nutr* **117**, 287–294.
38. Ribeiro MRC, da Silva AAM, de Brito MTSS *et al.* (2017) Effects of socioeconomic status and social support on violence against pregnant women: a structural equation modeling analysis. *PLoS One* **12**, e0170469.
39. Byrne BM (2013) *Structural Equation Modeling with Mplus: Basic Concepts, Applications, and Programming*. New York: Routledge.
40. Scaglioni S, Arrizza C, Vecchi F *et al.* (2011) Determinants of children's eating behavior. *Am J Clin Nutr* **94**, 2006S–2011S.
41. Anzman S, Rollins B & Birch L (2010) Parental influence on children's early eating environments and obesity risk: implications for prevention. *Int J Obes* **34**, 1116–1124.
42. Kral TVE & Rauh EM (2010) Eating behaviors of children in the context of their family environment. *Physiol Behav* **100**, 567–573.
43. DeBoer MD, Scharf RJ & Demmer RT (2013) Sugar-sweetened beverages and weight gain in 2-to 5-year-old children. *Pediatrics* **132**, 413–420.
44. Shao T, Tao H, Ni L *et al.* (2016) Maternal pre-pregnancy body mass index and gestational weight gain with preschool children's overweight and obesity. *Zhonghua Yu Fang Yi Xue Za Zhi* **50**, 123–128.
45. Li N, Liu E, Guo J *et al.* (2013) Maternal prepregnancy body mass index and gestational weight gain on offspring overweight in early infancy. *PLoS One* **8**, e77809.
46. Kakinami L, Barnett TA, Séguin L *et al.* (2015) Parenting style and obesity risk in children. *Prev Med* **75**, 18–22.
47. Wardle J, Carnell S, Haworth CMA *et al.* (2008) Evidence for a strong genetic influence on childhood adiposity despite the force of the obesogenic environment. *Am J Clin Nutr* **87**, 398–404.
48. Rooney K & Ozanne SE (2011) Maternal over-nutrition and offspring obesity predisposition: targets for preventative interventions. *Int J Obes* **35**, 883–890.
49. Avena NM, Rada P & Hoebel BG (2008). Evidence for sugar addiction: behavioral and neurochemical effects of intermittent, excessive sugar intake. *Neurosci Biobehav Rev* **32**, 20–39.
50. Westwater ML, Fletcher PC & Ziauddeen H (2016) Sugar addiction: the state of the science. *Eur J Nutr* **55**, 55–69.
51. Naska A, Bountziouka V, Trichopoulou A *et al.* (2010) Soft drinks: time trends and correlates in twenty-four European countries. A cross-national study using the DAFNE (Data Food Networking) databank. *Public Health Nutr* **13**, 1346–1355.
52. Monteiro CA, Levy RB, Claro RM *et al.* A new classification of foods based on the extent and purpose of their processing. *Cad Saude Publica* **26**, 2039–2049.
53. Levy RB, Claro RM, Mondini L *et al.* (2011) Regional and socioeconomic distribution of household food availability in Brazil, in 2008–2009. *Rev Saude Publica* **46**, 6–15.
54. Bielemann RM, Motta JVS, Minten GC *et al.* (2015) Consumption of ultra-processed foods and their impact on the diet of young adults. *Rev Saude Publica* **49**, 28.
55. Jaime PC, Delmuè DCC, Campello T *et al.* (2018) A look at the food and nutrition agenda over thirty years of the Unified Health System. *Ciênc saúde coletiva* [online] **23**, 1829–1836.
56. Clausen T, Slott M, Solvoll K *et al.* (2001) High intake of energy, sucrose, and polyunsaturated fatty acids is associated with increased risk of preeclampsia. *Am J Obstet Gynecol* **185**, 451–458.
57. Borgen I, Aamodt G, Harsem N *et al.* (2012) Maternal sugar consumption and risk of preeclampsia in nulliparous Norwegian women. *Eur J Clin Nutr* **66**, 920–925.
58. Englund-Ögge L, Brantsæter AL, Haugen M *et al.* (2012) Association between intake of artificially sweetened and sugar-sweetened beverages and preterm delivery: a large prospective cohort study. *Am J Clin Nutr* **96**, 552–529.
59. Petherick ES, Goran MI & Wright J (2014) Relationship between artificially sweetened and sugar-sweetened cola beverage consumption during pregnancy and preterm delivery in a multi-ethnic cohort: analysis of the Born in Bradford cohort study. *Eur J Clin Nutr* **68**, 404–407.
60. Muthén LK & Muthén B (2012) *Statistical Analysis with Latent Variables: Mplus User's Guide*. Los Angeles, CA: Muthén & Muthén.