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ABSTRACT

Background: Maternal characteristics and childhood growth have been identified as risk factors for eating disorders (ED). Most of the studies to date have been unable to investigate these factors prospectively while accounting for their interdependencies. We address this by investigating whether the association of maternal pre-pregnancy body mass index (ppBMI) with adolescent ED behaviors can be explained by childhood growth and/or a concurrent environmental pathway captured by maternal eating habits.

Methods: Data from girls participating in the Avon Longitudinal Study of Parents and Children (ALSPAC), a prospective cohort based in the UK, who had information on parentally and self-reported ED behaviors at age 13/14 years (n=3,529). Data also include: maternal ppBMI and eating habits when the child was 8 years old, child's birth weight, BMI from age 7 to 12, pubertal development at 11 years, and relevant confounders. The contributions of childhood growth and concomitant maternal eating habits to the association of maternal ppBMI with ED behaviors were quantified in terms of interventional disparity effects for multiple mediators.

Results: Maternal pre-pregnancy underweight was negatively associated with ED behaviors while overweight/obesity had the opposite relationship. Both were nearly fully explained by childhood growth.

Conclusions: Although maternal ppBMI is associated with developing ED, its role needs to be understood in the context of childhood factors, in particular childhood growth. The relatively small size of the remaining associations, once growth factors are hypothetically equalized across levels of maternal ppBMI, suggests that childhood growth is an important area for prevention.

Key words: ALSPAC, eating disorders, risk, mediation, interventional effects, disparity effects, maternal weight

Introduction

Eating disorders (ED) are chronic psychiatric illnesses comprising a range of conditions across the weight spectrum (anorexia nervosa, bulimia nervosa, binge eating disorder and other specified feeding and eating disorders). ED have a peak of onset in adolescence;¹ and they are prevalent amongst young people, affecting between 5-10% of adolescents girls.²⁻⁵ ED behaviors, mapping onto clinical diagnoses but not reaching thresholds for a clinical diagnosis in current diagnostic manuals, are common in young females and they predict adverse consequences, such as depression, anxiety disorders, and substance use.^{2,4,6}

ED are multifactorial in terms of their etiology.⁷ However, efforts to understand developmental risk for ED in the broader context of parental and child factors have been hampered by the lack of longitudinal studies both covering the whole developmental period, *and* adequately modeling the role of multiple risk factors and their interaction. Developmental risk factors do not exert their effect in a vacuum, but are often highly correlated and might operate through their effects on other factors. For instance, birth weight, childhood BMI and early puberty have been suggested as risk factors for ED and ED behaviors.⁸⁻¹¹ Similarly, post-pregnancy maternal BMI has been found to be prospectively associated with ED behaviors in adolescence and young adulthood.^{12,13} An extensive body of literature has investigated maternal weight status in pregnancy in relation to mental health outcomes in childhood and adolescence;¹⁴⁻¹⁶ no previous studies however have sought to model the joint effects of maternal weight status, infant/childhood weight and pubertal status on ED behaviors. A further possible mechanism via which maternal factors may be associated to ED behaviors involves childhood exposure to maternal eating and attitudes to food. The aim of this paper is to clarify these prospective associations and related risk pathways over time, as they may aid focusing preventative and early intervention efforts. We draw upon available longitudinal data collected prospectively over a 15-year span as part of the Avon Longitudinal Study of Parents and Children (ALSPAC) on maternal weight status, child's birth weight, BMI childhood trajectories, pubertal development, maternal eating habits, and ED behaviors

in early adolescence. ED behaviors were reported separately by the participants and by their parents and thus allow an assessment of the robustness of findings to differential sources of reporting error. We focused the study on participating girls, due to the higher prevalence of ED behaviors in early adolescence among girls and the differential patterns of ED behaviors across genders.^{17,18}

We investigated the extent to which the adjusted association between maternal pre-pregnancy weight status (underweight or overweight/obese) and adolescent ED behaviors would remain if the distributions (conditionally on confounders) of selected childhood variables were made to be the same as those of children whose mothers were normal weight. The childhood variables were chosen to represent growth and environmental pathways of risk, with their contribution to the adjusted pre-pregnancy BMI (ppBMI)-ED behaviors association quantified in terms of interventional disparity indirect effects.^{19,20} This approach has the advantage of not demanding a causal interpretation with respect to the exposure, maternal ppBMI, (hence avoiding its related pitfalls²¹), while still investigating possible pathways of interventions involving intermediate variables, as has been done by VanderWeele and Robinson with race as the exposure.²² As well as focusing on interventional effects for a single mediator (or a set of mediators considered *en bloc*), we also make use of the extension to multiple mediator settings proposed by Vansteelandt and Daniel²⁰ that allows multiple mediator-specific pathways to be compared without requiring an assumption of no unmeasured common causes of one mediator with another.

Methods

Participants

ALSPAC is a longitudinal, population-based, prospective study of women and their children. All pregnant women living in the geographical area of Avon, UK, expected to deliver between 1st April 1991 and 31st December 1992 were invited to participate in the study. All participating women gave informed and written consent. The ALSPAC study website

contains details of all the data that are available through a fully searchable data dictionary: <http://www.bris.ac.uk/alspac/researchers/data-access/data-dictionary>. A total of 14,541 pregnancies were enrolled, resulting in 14,062 live births and 13,617 singleton children who were alive at 1 year of age²³. Additional 713 children enrolled in the cohort at age 7 years are not included in these analyses due to missing data on maternal BMI by design.²³ A total of 10,135 children from the initial cohort were still followed up at the age 13-year wave. Further exclusions were due to non-response at this wave, leaving 7,078 respondents with ED behavior data, 3,529 of whom girls (Figure 1).

Figure 1 about here

Main Outcomes

Parentally-reported ED behaviors (p-ED: at mean child age 13.1 years (standard deviation, SD=0.2), data were collected via the Developmental and Well-being assessment (DAWBA), a semi-structured validated interview that generates a range of psychiatric diagnoses in children and adolescents.²⁴ The ED section of the DAWBA was given to parents and comprises 28 questions on ED behaviors and cognitions. These were used to derive three disordered eating patterns: 1. Binge eating/overeating; 2. shape and weight concern and weight control behaviors, and 3. food restriction, using exploratory structural equation modeling (as described in ¹⁷). Data on these three patterns were available on 3,529 girls. Because these are latent factors derived from structured questionnaires, they are standardized measures with mean 0 and standard deviation (SD) of 1.

Secondary Outcomes

Self-reported ED (s-ED) behaviors were obtained from the children at mean age 14.0 years (SD=0.2) using validated questions adapted from the Youth Risk Behavior Surveillance

System questionnaire²⁵ enquiring about the previous year; for details see ². We used two behaviors: binge eating, and fasting (which map closely onto the first and third the p-ED patterns). These self-reported outcomes were available for 2,751 and 2,734 girls, respectively.

Exposure

Maternal ppBMI, (kg/m²) was obtained from self-reported height and weight at enrolment during pregnancy, and used as an indicator of maternal weight status before the child's birth. It was categorized as: underweight (BMI<18.5), normal weight (18.5-24.9), overweight/obese (≥ 25) according to WHO criteria,²⁶ with normal weight treated as the reference category. Self-reported weight was highly correlated with maternal pregnancy objective weight in ALSPAC.²⁷

Mediators

Birth weight (g) was obtained from obstetric records. Childhood growth was quantified in terms of predicted random intercepts and slopes of the individual childhood trajectories of body mass index (BMI, kg/m²). These were derived from the original measurements taken at around age 7.5, 8.6, 9.8, 10.6, 11.8, and 12.8 years using a linear mixed effects model after log-transformation to achieve near normality. Assuming that the timing and frequency of the observations were unrelated to actual BMI values, the best fitting model had a linear and a quadratic term in age with random intercepts and random slopes for the linear age term only. Empirical Bayes predictions of the random intercepts and slopes were then saved and used to generate individual-level BMI at age 7.5 (thereafter labeled '*size*') and BMI rate of increase ('*yearly velocity*') (details in eTable 1). Pubertal development was defined using Tanner's stage of breast development at mean child age 10.7 years,²⁸ based on parental reports. This was categorized as early (Tanner stage ≥ 2) or age appropriate (< 2). At child age 8 years, mothers were sent a questionnaire asking about their own eating habits. Factor analyses revealed two dimensions: (i) avoidance of new foods, and (ii) poor enjoyment of eating

(Micali et al in preparation). These were correlated with maternal self-reported ED at enrolment.

Covariates

Several potential confounders for the exposure-mediator, exposure-outcome and mediator-outcome relationships, were considered. These included maternal education and age, and lowest parental social class, all obtained at enrolment.²³ At 12 weeks' gestation women were asked about any recent or past history of: severe depression, schizophrenia, alcoholism, anorexia nervosa, bulimia nervosa and other psychiatric disorders. Multiple answers were possible; therefore, women could report more than one disorder. This information was combined into a variable indicating presence of any pre-pregnancy psychopathology.

Statistical methods

Definitions of effects of interest

The aim of the study was to investigate the covariate-adjusted association between maternal pre-pregnancy weight status and offspring ED behaviors, and to investigate the extent to which it is explained via a “growth pathway” and a “maternal environmental pathway” (Figure 2). The growth pathway comprises pathways from ppBMI to ED that pass through birth weight, BMI size and velocity, and timing of puberty; the maternal environmental pathway comprises pathways that pass through the two latent dimensions measuring her eating habits.

Figure 2 about here

We defined the contributions of these pathways in terms of interventional disparity indirect effects, initially with all 6 mediators contributing to the growth and environmental pathways taken *en bloc*, then with the two groups of mediators taken separately. Interventional disparity

indirect effects are a variant on interventional indirect effects;^{19,20} they borrow an idea from the recent literature on counterfactual disparity measures,^{22,29} and are described below.

In the setting with a vector of mediators, interventional indirect effects (as defined by VanderWeele, Vansteelandt and Robins¹⁹) compare what, on average, would occur to the outcome had all individuals in the population had their mediators set to take random values from their joint distribution, conditional on confounders, among the exposed versus the corresponding distribution among the unexposed, conditional on confounders, while the exposure had been set to be exposed for all; thus it captures the effect of a hypothetical intervention that would shift the distribution of all mediators, whilst fixing the exposure. This definition is causal with respect to the effects of both the exposure and the mediators on the outcome, and thus a meaningful quantitative interpretation requires consideration of the nature of the entailed hypothetical interventions on the exposure and mediators. As has been widely discussed^{30,31} this is difficult (and would typically involve complex stochastic hypothetical interventions,³²⁻³⁶) especially for variables such as BMI. In this context, therefore, we do not seek a strict causal interpretation with respect to the exposure, and pursue an alternative specification following VanderWeele and Robinson,^{22,37} and Naimi et al.²⁹ The *interventional disparity measure* indirect effects we define here pertain to the extent by which ED behaviors of girls whose mothers were underweight (or overweight/obese) before pregnancy would change, had the distributions of their mediators been changed to that of girls whose mothers were normal weight (conditional on confounders). Their complement, the direct effects, represent the covariate-adjusted associations (between ppBMI and ED behaviors) that would remain if all mediators were set to have the same joint distribution, given confounders, as is actually the case amongst girls whose mothers were normal weight.

For completeness and clarity, we write the effects mathematically below. Let X be the exposure (ppBMI), \mathbf{M} the vector of all six mediators, Y the outcome (ED behavior score), and \mathbf{C} the vector of four possible confounders. Write $Y(\mathbf{m})$ to be the potential value that Y would

take if \mathbf{M} were intervened upon and set to level \mathbf{m} . Let $\mathbf{M}^x\mathbf{c}$ be a random draw from the joint distribution of \mathbf{M} given \mathbf{C} among those with $X=x$. The Interventional Disparity Measure (IDM) direct and indirect effects, or IDM-DE and IDM-IE, are defined as follows, for categorical \mathbf{C} (with corresponding integrals and densities for continuous \mathbf{C}):

$$\text{IDM-DE} = \sum_{\mathbf{c}} [\text{E}\{Y(\mathbf{M}^0\mathbf{c})|X=1, \mathbf{C}=\mathbf{c}\} - \text{E}\{Y(\mathbf{M}^0\mathbf{c})|X=0, \mathbf{C}=\mathbf{c}\}] \text{Pr}(\mathbf{C}=\mathbf{c}),$$

$$\text{IDM-IE} = \sum_{\mathbf{c}} [\text{E}\{Y(\mathbf{M}^1\mathbf{c})|X=1, \mathbf{C}=\mathbf{c}\} - \text{E}\{Y(\mathbf{M}^0\mathbf{c})|X=1, \mathbf{C}=\mathbf{c}\}] \text{Pr}(\mathbf{C}=\mathbf{c}).$$

Note that these differ from the definitions given by VanderWeele³⁷ only to the extent that we marginalize over the distribution of covariates \mathbf{C} .

Under the identifying assumptions described in the next section the sum of the IDM-DE and IDM-IE is the \mathbf{C} -adjusted marginal association between X and Y expressed as a mean difference, which we label the *adjusted total association*, Adj-TA. That is,

$$\text{IDM-DE} + \text{IDM-IE} = \text{Adj-TA} = \sum_{\mathbf{c}} \{\text{E}(Y|X=1, \mathbf{C}=\mathbf{c}) - \text{E}(Y|X=0, \mathbf{C}=\mathbf{c})\} \text{Pr}(\mathbf{C}=\mathbf{c}).$$

Assumptions

The identification of the above effects relies on a number of assumptions, commonly referred to as ‘no interference’, consistency, and ‘no unmeasured confounding’. Due to our focus on effects that avoid a causal interpretation with respect to the exposure (ppBMI), the precise nature of these assumptions is somewhat different (weaker) than usually stated (and furthermore do not require a cross-world independence or similar assumption). In the present context, the assumption of no interference states that the ED behavior of one girl is not influenced by the mediator levels of another; and the assumption of consistency states that within a group of girls, all of whom share the same mediator levels, \mathbf{m} say, the same background confounder levels \mathbf{c} and the same exposure level x , the mean ED behavior level in this group were we hypothetically to intervene and set their mediator values to \mathbf{m} would be

the same as the actual mean ED behaviour level in this group; this should be true at all possible levels of confounders, exposure and mediators. Written mathematically:

$$E\{Y(\mathbf{m})|C=\mathbf{c},X=x,M=\mathbf{m}\} = E(Y|C=\mathbf{c},X=x,M=\mathbf{m}), \text{ for all } \mathbf{c},x,\mathbf{m}.$$

This version of the consistency assumption for mediation analysis is weaker than usually stated, still may not be met in applications, as expanded in the discussion. Finally, no unmeasured confounding in the present context states that the potential ED behavior score were the mediators set by hypothetical intervention to a particular set of levels should be conditionally mean independent of the actual mediator levels, conditional on exposure and confounders; this should be true at all possible levels of confounders, exposure and mediators. Hence, unmeasured exposure-mediators and exposure-outcome common causes are permitted (indicated by V and W in Figure 2). This is the rigorous way of saying that there can be no unmeasured mediator-outcome confounding; written mathematically:

$$E\{Y(\mathbf{m})|C=\mathbf{c},X=x,M=\mathbf{m}\} = E(Y(\mathbf{m})|C=\mathbf{c},X=x), \text{ for all } \mathbf{c},x,\mathbf{m}.$$

Vansteelandt and Daniel²⁰ extend the definition of interventional effects to multiple mediators, and allow for the partitioning of the indirect effect into effects that involve subsets of the mediators, plus an additional indirect effect representing the dependence between mediators. The mediators in Vansteelandt and Daniel's²⁰ formulation are permitted to be correlated via factors that are unmeasured (indicated by U in Figure 2); this, together with our focus again on disparity measure effects (in contrast to Vansteelandt and Daniel) means that no additional assumptions from those stated above are needed for our investigation of separate indirect effects through subsets of multiple mediators. We adopted this approach when separating the mediators into growth and maternal environment subsets. The precise definitions of these interventional disparity measure multiple mediator effects can be found in the e-Appendix.

Note that if we additionally made assumptions that justified a causal interpretation of Adj-TA, then the sum of IDM-DE and IDM-IE would represent the total causal effect of X on Y expressed as a marginal mean difference: $E\{Y(1) - Y(0)\}$. Even without these additional assumptions, the decomposition of the adjusted total association is meaningful, as it allows the examination of alternative pathways.

Estimation method

Estimation was via a series of richly-specified regression models, combined using Monte Carlo simulation performed using Stata v14.2.³⁸ This required the specification of parametric models for the outcome given exposure, mediators and confounders, and for the mediators given exposure and confounders, on a 1000-fold expanded dataset. By richly-specified, we mean that many interactions and other higher-order terms were included, in an attempt to lessen the impact on the final estimates of incorrectly-specified parametric models. Full details can be found in the e-Appendix. Standard errors were estimated using the non-parametric bootstrap (with 1000 bootstrap samples) and used to calculate 95% confidence intervals. All mediated effects are expressed as mean differences, and thus when the outcome is binary (s-ED), these are risk differences.

Missing data

Data on exposure, confounders and mediators were affected by missingness. For this reason, single stochastic imputation using chained equations³⁹ with 10 burn-in iterations was implemented before the 1000-fold data expansion for the Monte Carlo estimation procedure was carried out, under the assumption that missingness was at random (MAR).⁴⁰ In this instance, this implies that common drivers of missingness and the partially-observed variables are included among the variables being conditioned upon in the imputation. The imputation models were all more general than the analytical models. Multiple imputation was not

required since the bootstrap was used to estimate SEs, and the imputation step was re-done on each bootstrap sample.

Results

Data on p-ED behaviors were available on 3,529 girls (Figure 1). A comparison of baseline characteristics of these girls against those included in the ALSPAC study at birth shows some attrition linked to maternal education (eTable 2). Among the girls included in this study, and who represent 70% of those that were invited to participate at age 13 years (mothers of 1,562 girls -30% of those invited- did not return questionnaires), exposure, mediators and confounders were affected by missingness, with 1,989 (56%) having complete information on all relevant variables. Missingness was associated with younger maternal age, lower education, parental manual social class, lower birth weight and greater childhood BMI. The subset with complete records had slightly lower scores of p-ED behaviors and a slightly lower prevalence of s-ED behaviors (Table 1 and eTable 3).

[Table 1 about here](#)

Separate adjusted associations between each outcome and each of the exposure and mediators are shown in Table 2. Maternal pre-pregnancy weight status and childhood variables were in general strongly associated with ED behaviors (Table 2). Much weaker, but similar, associations are seen for s-ED (Table 2). There was no evidence of associations between any of the outcomes and maternal eating habits (Table 2). Further exploration of associations between each potential mediator and the exposure shows strong and consistent relationships between each mediator and maternal ppBMI, except for maternal avoidance of new foods (Table 3).

[Table 2 and Table 3 about here](#)

Table 4 reports the estimated adjusted total association between ppBMI and p-ED behaviors and their partitioning into interventional disparity direct and indirect effects, all expressed as mean differences in p-ED behaviors scores and obtained using the full set of 3,529 girls. The estimated adjusted total association comparing maternal underweight vs normal on binge eating/overeating was negative (-0.18, 95% confidence interval (CI): -0.29, -0.06) and of similar magnitude to the estimated interventional indirect effect via the six mediators taken *en bloc* (-0.22, 95% CI: -0.32, -0.11).

Adj-TA=-0.18 (95% CI: -0.29, -0.06) represents the strength of association between maternal pre-pregnancy underweight and binge eating/overeating, estimated after adjusting for (and then standardising by) the baseline covariates (maternal age, education, and psychopathology, and parental social class). The estimated DM-DE=0.04 (95% CI: -0.09, 0.17) represents the extent of this adjusted association that would remain if the six mediators were set to have the same distribution in girls whose mothers were underweight before pregnancy, as that of girls whose mothers were normal weight (conditional on confounders). By complement, the estimated DM-IE=-0.22 (95% CI: -0.32, -0.11) represents the extent by which the ED score of girls whose mothers were underweight before pregnancy would change, if the six mediators taken *en bloc* were set to have the same distribution as that of girls whose mothers were normal weight (conditional on confounders).

Similar differences and decompositions were found for weight and shape concern & weight control behaviors and food restriction.

The Adj-TA of maternal overweight/obesity vs. normal for each p-ED behavior was in the opposite direction to that for maternal underweight (binge eating/overeating: 0.25 (0.18, 0.32); weight and shape concern & weight control behaviors: 0.22 (0.15, 0.29); food restriction: 0.18 (0.11, 0.25); Table 4). These effects were fully explained by the 6 mediators when taken *en bloc*.

[Table 4 about here](#)

When the 6 mediators were split into a “growth pathway” (captured by birth weight, childhood growth and puberty status), and a “maternal environmental pathway” (captured by the two dimensions of maternal eating habits), we found that the first pathway explained most of the indirect effect of maternal ppBMI on ED behaviors (Table 4).

If the assumptions discussed in the Methods are deemed to be met, these estimates of indirect effects via the growth pathway quantify the extent by which the ED scores of girls whose mothers were underweight (or overweight/obese) would change if the distributions of the childhood growth variables (but not those of maternal eating habits) were made to be the same as those of children whose mothers were normal weight (conditionally on confounders).

eTable 4 reports risk differences for the two s-ED behaviors. The estimated effects for maternal overweight are in line with those from the parental reports although they are less precise. Those for the effects of maternal underweight on self-reported fasting instead indicate that the protective association is not explained by any of the mediators considered here. All the findings are consistent with the adjusted relationships observed in the data, in particular with the strength of association with the mediators belonging to the growth pathway.

Discussion

Parental and developmental risk factors for childhood disorders have often been studied independently; however most developmental risk factors, especially weight, growth and parental weight status are highly associated. Therefore, studying them independently might not provide a full account of risk pathways. We provide evidence of the importance of studying related intergenerational risk factors (in particular maternal weight status, child weight, growth and pubertal development) for adolescent ED using a causal inference framework. Existing evidence suggests that child weight, growth and parental weight might be predictors of ED.⁸⁻¹³ However, few large comprehensive prospective studies are available,

therefore no studies (to our knowledge) have investigated how these factors might be related, nor relevant intergenerational risk pathways.

This study is the first to show a differential (protective vs. risk-conferring) adjusted association between pre-pregnancy maternal underweight vs normal and overweight/obesity vs normal and adolescent ED behaviors. We found that these adjusted associations were almost fully explained by a growth pathway (with a strong biological component) involving the child's birth weight, growth and early puberty. Shifting from the role of individual risk factors to a broader perspective which includes risk pathways has the potential not only to improve our understanding of the role of intergenerational risk for ED, but also, potentially, to target our prevention and early intervention efforts where they might be more effective. Secondly, given the increasing evidence of the importance of obesity genetic risk for eating behaviors and weight development,^{42,43} new evidence on how biological risk pathways affect eating and ED is likely to influence novel conceptualizations of the pathophysiology of ED and eating development.

Our findings need to be understood in the context of the strengths and limitations. The data comprises information collected prospectively over 15 years as part of the ALSPAC Study. This birth cohort suffers from attrition linked to socio-economic status as also noted by Howe *et al.*⁴¹. These authors found that even an attrition of up to 50%, which was observed at age 15 years, did not affect the qualitative conclusions drawn from ALSPAC on the association between social inequalities and several outcomes when based on crude analyses of the complete records. We have used ALSPAC data up to age 13 years and controlled in our analyses for the main drivers of attrition, namely socio-economic indicators. Furthermore, by imputing the variables affected by missingness, which pattern was found to be influenced too by socio-economic factors, and by controlling for them in the analyses, the likely bias due to attrition and item-response missingness should have been reduced, if the assumption that missingness and attrition are at random, given the observed data, is justified.

The majority of data were collected objectively. Moreover, we used novel approaches to mediation analysis to try and distinguish pathways along which maternal BMI may be associated with the outcome, distinguishing between a growth and a more environmental component that allowed for unmeasured common causes of their distributions. However, these analyses rely on strong unverifiable assumptions besides MAR, namely no unmeasured confounding of the mediator-outcome relationships, no interference and consistency (again for the mediator-outcome relationships). To attempt to meet the first of these assumptions we have controlled for likely confounders, including two indicators of socio-economic position that may capture, at least in part, the effect of other unmeasured confounders. No interference would not be satisfied for example, if the eating habits of a girl's mother influenced the ED of another girl, as might occur if they regularly socialised with each other's families. The ALSPAC participants however are located in a fairly wide geographical area and so we can plausibly assume that this would affect only a minority. It is implausible that a single (simple) hypothetical intervention on the mediators exists that would lead to the consistency assumption being satisfied, especially for those involved in the growth pathway. For example, there are many different hypothetical ways of 'setting' the growth trajectory of girls, and each may lead to a different ED behavior level; furthermore, our dataset will contain girls who attain their growth trajectory for many different reasons. The consistency assumption thus necessitates that we interpret our effects in terms of a complex hypothetical intervention, which randomly assigns girls to have their growth trajectory set in one of many different ways, such that the overall intervention is 'non-invasive' in the sense that it would not change the outcome for those whose mediators are being 'set' to the same value as was in fact attained. For further discussion of these issues, see ^{32,33,35}.

Although our main analyses focused on parentally-reported ED behaviors, we also replicated our analyses on self-reported ED behaviors, showing consistency of our results. The main limitations entail the nature of the sample, representative of a selected (by attrition and by design, as only pregnant women were included in the study) UK population, but limited in its

generalizability to other populations. Our exposure, maternal ppBMI, was based on self-report; however, using questionnaires rather than objective measures is cost-effective in the context of large samples and maternal self-reported weight in this sample was highly correlated with objectively measured weight.²⁷ Maternal underweight was not highly prevalent (~5%), leading to imprecision of our estimates of effects comparing maternal underweight versus normal weight. Although we were not able to study maternal ED as an independent predictor, they were included among the confounders as a component in maternal pre-pregnancy mental health disorders; however, the overall prevalence was low and we acknowledge this is an imperfect measure of maternal psychopathology. It is plausible that a subset of women who were underweight might have suffered from restrictive eating disorders. In relation to our outcomes, our aim was to focus on ED behaviors that are prevalent in the community,²⁻⁴ rather than full-blown ED (rarer at the developmental stage under investigation). However future studies will aim to determine whether similar risk mechanisms are at play in ED.

In conclusion, this study highlights the importance of examining intergenerational effects using comprehensive explanatory models that avoid focusing on specific variables in a vacuum. We confirmed our hypothesis that maternal ppBMI is conditionally associated with child eating behaviors, and that the majority of this adjusted association acts through a pathway driven by birth weight, growth and puberty. Future studies should extend this investigation to specific genetic or metabolic risk.

Table 1 - Means and standard deviations, or frequencies and percentages (*italics*), of main variables in the whole study and in the complete records ^a subset

	Overall			Complete records		
	N	Mean/Freq	SD/%	N	Mean/Freq	SD/%
<i>Outcomes</i>						
<u>Parental report</u>						
Binge eating/ overeating (SD) ^b	3,529	0.00	1.00	1,989	-0.04	0.98
Weight concern/control (SD) ^b	3,529	0.00	1.00	1,989	-0.03	0.99
Food restriction (SD) ^b	3,529	0.00	1.00	1,989	-0.03	0.97
<u>Self-report</u> ^a						
Binge eating						
Yes	2,751			1,679		
No		188	6.8		99	5.9
		2,563	93.2		1,580	94.1
Fasting						
Yes	2,734			1,662		
No		247	9.0		130	5.9
		2,487	91.0		1,532	92.2
<i>Exposure</i>						
PP-maternal weight status						
Underweight	3,088			1,989		
Normal weight (reference group)		154	5.0		92	4.6
Overweight/Obese		2,344	75.9		1,529	76.9
		590	19.1		368	18.5
<i>Mediators</i>						
Birth weight (SD) ^b	3,330	0.06	0.96	1,989	0.09	0.90
BMI size at 7y (SD) ^{b,c}	3,238	0.00	1.00	1,989	-0.05	0.97
BMI yearly velocity (7-12y) (SD) ^{b,c}	3,238	0.00	1.00	1,989	0.03	1.00
Pubertal development at age 12y						
Age appropriate		2,138	77.0		1,529	76.9
Early		638	23.0		460	23.1
Maternal avoidance of new foods (8y) (SD)	2,942	-0.02	0.99	1,989	-0.03	0.99
Poor enjoyment of eating (8y) (SD)	2,942	-0.02	1.00	1,989	-0.05	0.97
<i>Confounders</i>						
Maternal age (y)						
<25	3,369			1,989		
25-29		521	15.4		239	12.0
≥30		1,336	39.7		811	40.8
		1,512	44.9		939	45.2
Parental social class						
Manual/low	3,091			1,989		
Non-manual/high		428	13.9		234	11.8
		2,663	86.1		1,755	88.2
Maternal education						
Up to secondary	3,185			1,989		
Secondary or higher		1,772	55.6		1,019	48.9
		1,413	44.4		970	51.2
Maternal lifetime psychopathology reported in pregnancy						
None reported	3,250			1,989		
Any		2,880	88.6		1,801	90.6
		370	11.4		188	9.5

N: records with information; Freq: frequency; SD: standard deviation; y: years.

a: The definition of complete records did not include self-reported ED

b: Internally standardized (before exclusions), with units expressed in terms of SDs.

c: Predicted values from a mixed effects model fitted to the repeated childhood BMI measures.

Table 2 – Estimated regression coefficients (β) or odds ratios (ORs) for associations between ED behaviors (dependent variable, internally standardized) and, separately, exposure and mediators, adjusted for relevant confounders ^a

ED behaviors (parentally-reported)	N	Binge eating /overeating ^b		Weight and shape concern & weight control behaviors ^b		Food restriction ^b	
		β	95% CI	β	95% CI	β	95% CI
Exposure							
Pp-maternal weight status	2,874						
Underweight		-0.14	-0.31, 0.03	-0.20	-0.37, -0.03	-0.19	-0.36, -0.02
Normal weight (ref. group)		0		0	-	0	-
Overweight/Obese		0.31	0.22, 0.40	0.26	0.17, 0.35	0.22	0.13, 0.31
<i>Linear trend (p-value)</i>		<0.001		<0.001		<0.001	
Mediators							
Birth weight (SD)^b	2,840	0.06	0.02, 0.10	0.05	0.01, 0.10	0.04	0.00, 0.08
BMI size at 7y (SD)^{b,c}	2,675	0.33	0.29, 0.36	0.33	0.29, 0.36	0.27	0.23, 0.31
BMI yearly velocity (7-12y) (SD)^{b,c}	2,675	0.11	0.07, 0.15	0.15	0.11, 0.18	0.13	0.09, 0.17
Pubertal development (12y)	2,330						
Age-appropriate		Ref		Ref		Ref	
Early		0.37	0.28, 0.46	0.44	0.34, 0.53	0.44	0.34, 0.53
Maternal avoidance of new foods (8y) (SD)	2,499	0.01	-0.03, 0.05	0.02	-0.02, 0.06	0.02	-0.02, 0.06
Maternal poor enjoyment of eating (8y)(SD)	2,499	-0.01	-0.04, 0.04	0.01	-0.03, 0.05	0.02	-0.02, 0.06

ED behaviors (self-reported)	N	Binge eating		N	Fasting	
		OR	95% CI		OR	95% CI
Exposure						
PP-maternal weight status	2,279			2,236		
Underweight		0.68	0.27, 1.71		0.94	0.46, 1.91
Normal weight (ref. group)		1	-		1	-
Overweight/obese		1.08	0.71, 1.47		1.37	0.96, 1.96
<i>Linear trend (p-value)</i>		0.45			0.10	
Mediators						
Birth weight (SD)^b	2,230	1.10	0.90, 1.33	2,210	0.93	0.79, 1.09
BMI size at 7y (SD)^{b,c}	2,147	1.49	1.25, 1.77	2,110	1.67	1.43, 1.59
BMI yearly velocity (7-12y) (SD)^{b,c}	2,147	0.98	0.82, 1.16	2,110	0.97	0.83, 1.13
Pubertal development at 12y	2,037			1,997		
Age-appropriate		Ref	-		Ref	-
Early		1.27	0.85, 1.89		1.86	1.32, 2.61
Maternal avoidance of new foods (8y) (SD)	2,036	0.77	0.63, 0.94	2,001	0.84	0.71, 1.00
Maternal poor enjoyment of eating (8y) (SD)	2,036	0.84	0.67, 1.04	2,001	1.12	0.96, 1.30

^a Estimates were adjusted as follows:

- For the exposure: parental social class, maternal education, age and psychopathology
- For the mediators: as above plus pp-BMI

^b Internally standardized (before exclusions), with units expressed in terms of SDs.

^c Predicted values from a mixed effects model fitted to the repeated childhood BMI measures.

Table 3 – Estimated regression coefficients or odds ratios (OR, *in italics*) for the association between each mediator (dependent variable) and pp-BMI

	N	Pre-pregnancy maternal weight status (pp-BMI)					<i>Trend</i> (<i>p</i> -value)
		Underweight		Normal weight	Overweight/obese		
		Regression coeff./OR	95% CI		Regression coeff./ OR	95% CI	
<i>Mediators</i>							
Birth weight (SD)^a	2,840	-0.33	-0.49, -0.18	Ref.	0.24	0.16, 0.33	<0.001
BMI size at 7y (SD) ^{a,b}	2,675	-0.30	-0.47, -0.13	Ref.	0.51	0.42, 0.61	<0.001
BMI yearly velocity (7-12y) (SD)^{a,b}	2,675	-0.19	-0.36, -0.01	Ref.	0.12	0.02, 0.21	0.001
<i>Pubertal stage at age 12y^c</i>	2,330	<i>0.49</i>	<i>0.27, 0.89</i>	Ref.	<i>1.71</i>	<i>1.35, 2.16</i>	<0.001
Maternal avoidance of new foods (8y) (SD)	2,499	0.06	-0.13, 0.24	Ref.	-0.03	-0.13, 0.07	0.43
Maternal poor enjoyment of eating (8y) (SD)	2,499	0.41	0.23, 0.59	Ref.	-0.16	-0.25, -0.06	<0.001

^a Estimates were adjusted for parental social class, maternal education, age and psychopathology

^b Internally standardized (before exclusions), with units expressed in terms of SDs.

^c Predicted values from a mixed effects model fitted to the repeated childhood BMI measures (details in eTable 1).

Table 4– Adjusted total association of maternal pre-pregnancy BMI and ED behaviors and interventional disparity direct and indirect effects estimated by Monte Carlo simulation and imputation of missing values (SEs estimated using 1000 bootstrap samples); N=3,529; Monte Carlo sample of 3,529,000.

Outcome	Effect (all direct and indirect effects are IDM)	Maternal weight status (<i>pp</i> -BMI; reference: normal weight)			
		Underweight		Overweight/Obese	
		Mean difference	95% CI	Mean difference	95% CI
Binge eating /Overeating	Adjusted total association	-0.18	-0.29, -0.06	0.25	0.18, 0.32
	Direct	0.04	-0.09, 0.17	-0.02	-0.08, 0.05
	Indirect via all 6 mediators	-0.22	-0.32, -0.11	0.26	0.21, 0.32
	Indirect via “growth pathway” ^a	-0.22	-0.32, -0.11	0.28	0.23, 0.33
	Indirect via “maternal environment pathway” ^a	-0.01	-0.04, 0.03	-0.02	-0.04, -0.01
	Indirect via dependence of growth/puberty on maternal eating habits ^b	0.00	-0.01, 0.01	0.00	-0.01, 0.01
	Weight and shape concern & weight control behaviors	Adjusted total association	-0.20	-0.32, -0.07	0.22
Direct	0.08	-0.06, 0.22	-0.03	-0.10, 0.04	
Indirect via all 6 mediators	-0.28	-0.39, -0.17	0.25	0.20, 0.30	
Indirect via “growth pathway” ^a	-0.28	-0.39, -0.17	0.26	0.22, 0.31	
Indirect via “maternal environment pathway” ^a	0.00	-0.04, 0.04	-0.02	-0.03, 0.00	
Indirect via dependence of growth/puberty on maternal eating habits ^b	0.00	-0.01, 0.01	0.00	-0.01, 0.01	
Food Restriction	Adjusted total association	-0.21	-0.33, -0.09	0.18	0.11, 0.25
	Direct	0.01	-0.12, 0.15	-0.03	-0.10, 0.05
	Indirect via all 6 mediators	-0.22	-0.33, -0.11	0.20	0.16, 0.25
	Indirect via “growth pathway” ^a	-0.23	-0.33,-0.11	0.20	0.16, 0.25
	Indirect via “maternal environment pathway” ^a	-0.01	-0.03, 0.04	-0.01	-0.02, 0.01
	Indirect via dependence of growth/puberty on maternal eating habits ^b	0.00	-0.01, 0.01	0.01	-0.002, 0.01

IDM: Interventional disparity measure

^a The “growth pathway” involves birth weight, growth and puberty; the “maternal environment pathway” involves the two latent classes measuring attitude to food when the child was 8 year old.

^b This component represents the dependence between the two multivariate pathways in their indirect effects.

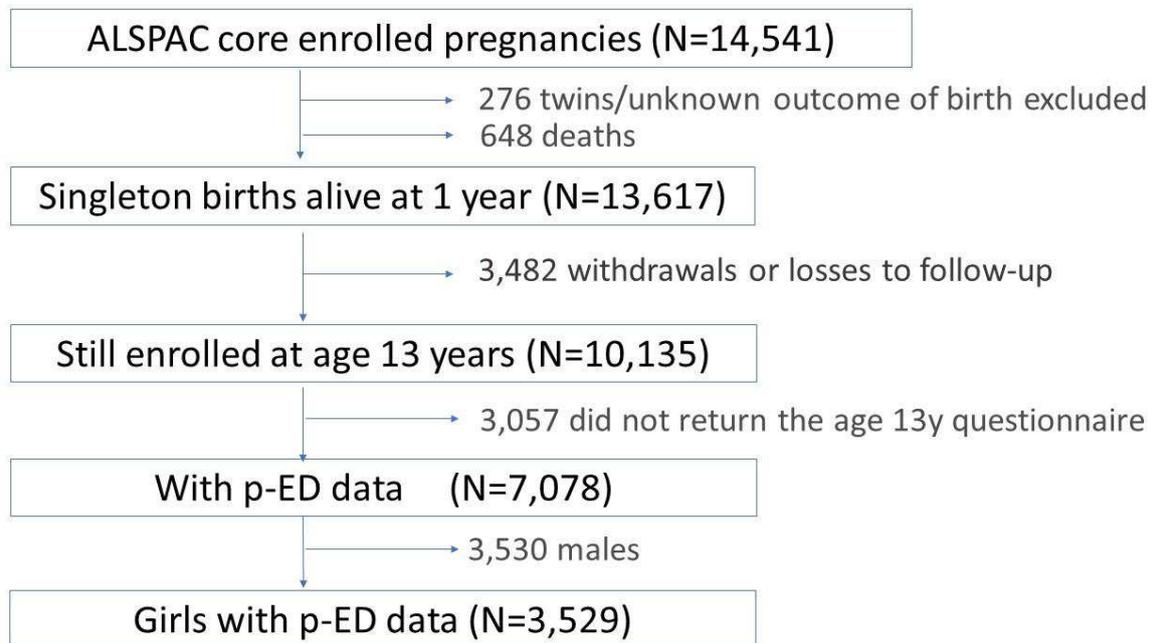


Figure 1: The study flow diagram

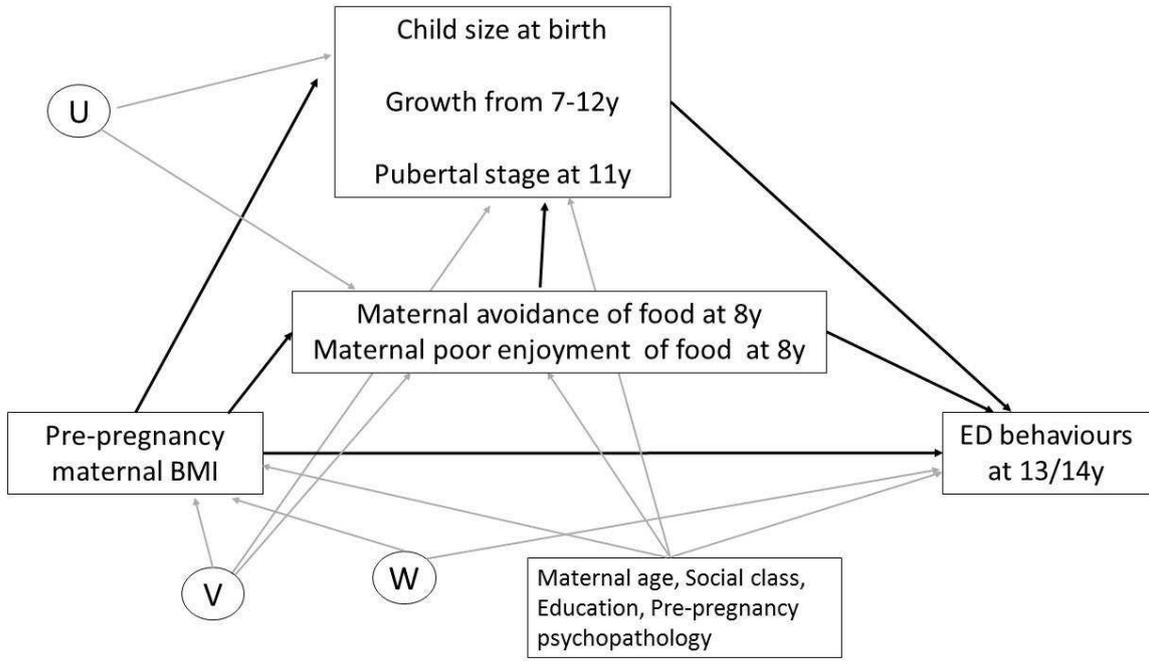


Figure 2: Presumed causal model

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