Original Article

What life course theoretical models best explain the relationship between exposure to childhood adversity and psychopathology symptoms: recency, accumulation, or sensitive periods?

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Background. Although childhood adversity is a potent determinant of psychopathology, relatively little is known about how the characteristics of adversity exposure, including its developmental timing or duration, influence subsequent mental health outcomes. This study compared three models from life course theory (recency, accumulation, sensitive period) to determine which one(s) best explained this relationship.

Methods. Prospective data came from the Avon Longitudinal Study of Parents and Children ($n = 7476$). Four adversities commonly linked to psychopathology (caregiver physical/emotional abuse; sexual/physical abuse; financial stress; parent legal problems) were measured repeatedly from birth to age 8. Using a statistical modeling approach grounded in least angle regression, we determined the theoretical model(s) explaining the most variability (r^2) in psychopathology symptoms measured at age 8 using the Strengths and Difficulties Questionnaire and evaluated the magnitude of each association.

Results. *Recency* was the best fitting theoretical model for the effect of physical/sexual abuse (girls $r^2 = 2.35\%$; boys $r^2 = 1.68\%$). Both *recency* (girls $r^2 = 1.55\%$) and *accumulation* (boys $r^2 = 1.71\%$) were the best fitting models for caregiver physical/emotional abuse. *Sensitive period* models were chosen alone (parent legal problems in boys $r^2 = 0.29\%$) and with *accumulation* (financial stress in girls $r^2 = 3.08\%$) more rarely. Substantial effect sizes were observed (standardized mean differences = 0.22–1.18).

Conclusions. Child psychopathology symptoms are primarily explained by recency and accumulation models. Evidence for sensitive periods did not emerge strongly in these data. These findings underscore the need to measure the characteristics of adversity, which can aid in understanding disease mechanisms and determining how best to reduce the consequences of exposure to adversity.

Introduction

One of the most consistent findings in psychiatric epidemiology is that childhood adversity, including maltreatment and stressful life events, is one of the most potent determinants of mental health problems throughout the lifespan (Shonkoff & Garner, 2012). Overall, childhood adversities appear to at least double the risk of youth- and adult-onset mental disorders (McLaughlin *et al.* 2010; McLaughlin *et al.* 2012; Gilman *et al.* 2015). Yet, relatively little is known about how the characteristics of adversity influence subsequent mental health outcomes. For instance, does the developmental timing of exposure to adversity matter most in shaping future risk for psychopathology symptoms? Or is the duration of exposure more important? A greater understanding of how the features of adversity are associated with mental health outcomes could shed new light on the mechanisms underlying risk for psychopathology, by suggesting developmental processes that are disrupted by exposure. It could also help in determining the optimal times to intervene, as childhood spans multiple developmental periods when different types of interventions (e.g., home- *v.* school-based programs) could

be deployed to minimize the effects of adversity based on the age of the child or the nature of the exposure.

Here, we compared three theoretical models derived from life course theory, each of which describes the association between an exposure and a health outcome (Ben-Shlomo & Kuh, 2002; Kuh & Ben-Shlomo, 2004), to determine the model(s) that best explained the relationship between exposure to childhood adversity on emotional and behavioral problems at age 8. The first life course model tested was an *accumulation of risk model*, which posits that every additional year of exposure is associated with an increased risk of poor health in a dose-response manner, irrespective of timing (Rutter *et al.* 1979; Evans *et al.* 2013). The second model was a *sensitive period model*, which presumes the developmental timing of exposure is most important. In this model, timing matters because the exposure occurrence coincides with the time period of greatest maturation or plasticity in the brain, for example (Bailey *et al.* 2001; Knudsen, 2004), making the exposure at one point in time more potent than the same exposure occurring earlier or later (Dunn *et al.* 2013; 2016). The third model was a *recency model*, which suggests that mental health outcomes are most strongly linked to more proximal, rather than distal events, as the effects of adversity can be time-limited (Shanahan *et al.* 2011). To our knowledge, no studies have simultaneously conducted formal comparisons of these three theoretical models across the main types of adversity related to psychopathology.

We aimed to address this gap by using an innovative life course modeling approach (Mishra *et al.* 2009) to simultaneously compare these theoretical models with four of the main types of early life adversity linked to psychopathology: caregiver physical or emotional abuse, sexual or physical abuse, financial stress, and parent legal problems. These adversities were measured repeatedly between birth and age 8. Our goal was to determine which theoretical model (or set of models) were best supported by the data, estimate the magnitude of the association between each model and child psychopathology symptoms, and evaluate whether the model chosen varied by the type of exposure. We performed these analyses separately among boys and girls, as prior studies have shown sex differences in lifetime exposure to adversity (Koenen *et al.* 2010) and risk for psychopathology (Dunn *et al.* 2012). Although these life course models are often described in relation to adult outcomes, and the period of childhood is often considered a sensitive period in and of itself, we focused on child psychopathology symptoms in order to examine the short-term consequences of adversity and determine the possibility of being able to differentiate between these life course models for early-onset psychopathology symptoms.

Methods

Sample and procedures

Data came from a prospective, longitudinal birth-cohort of children (Avon Longitudinal Study of Parents and Children; ALSPAC, Boyd *et al.* 2013). ALSPAC sampled children born to mothers living in the county of Avon, England (120 miles west of London) with estimated delivery dates between April 1991 and December 1992. Approximately 85% of eligible pregnant women agreed to participate ($n = 14\,541$), and 99% of eligible live births ($n = 14\,775$) who were alive at 12 months of age ($n = 14\,701$ children) were enrolled. Response rates have been good (75% completed at least one follow-up). More details are available on the ALSPAC website including a fully searchable data dictionary: http://www.bristol.ac.uk/alspac/researchers/access/. Ethical approval for the study was obtained from the ALSPAC Ethics and Law Committee and the Local Research Ethics Committee.

Measures

Exposure to abuse and stress

We examined four types of adversity measured using parent-mailed questionnaires. Each adversity was measured on at least five occasions before age 8 (see Table 2), with each measurement occasion analyzed separately due to different assessment time periods. The adversity types selected are commonly used to define 'early life adversity' (Felitti *et al.* 1998; Slopen *et al.* 2012; Slopen *et al.* 2014). The abuse-related variables were chosen because they aligned with previous work demonstrating the strong association between physical, sexual, and emotional abuse and subsequent mental disorders (Maniglio, 2009; Norman *et al.* 2012). Similarly, the stress-related variables were chosen based on previous work linking parental incarceration (Murray & Murray, 2010, Turney, 2014) and financial stress (Evans, 2004) to risk for psychopathology.

Abuse. *Caregiver physical or emotional abuse*: Children were coded as having been exposed to physical or emotional abuse if the mother, partner, or both responded affirmatively to any of the following items: (1) your partner was physically cruel to your children; (2) you were physically cruel to your children; (3) your partner was emotionally cruel to your children; (4) you were emotionally cruel to your children.

Sexual or physical abuse: Exposure to sexual or physical abuse was determined through an item asking the mother to indicate whether or not the child had been exposed to either sexual or physical abuse from anyone.

Stress. *Financial stress*: Mothers indicated using a Likert-type scale (1 = not difficult; 2 = slightly difficult; 3 = fairly difficult; 4 = very difficult) the extent to which the family had difficulty affording the following: (a) items for the child; (b) rent or mortgage; (c) heating; (d) clothing; (e) food. Children were coded as exposed if their mothers reported at least slight difficulty for three or more items; this cut-point roughly corresponded to the top quartile.

Parent legal problems: Mothers indicated whether or not the child's parents had been in trouble with the law in the past year. Children were coded as exposed if either or both parents had legal problems.

For each type of adversity, we generated three sets of encoded variables, as summarized in Table 1. These encoded variables were all entered into a multiple regression model for a given type of adversity, allowing for multiple life course associations to be present simultaneously.

As no clear sensitive periods linking exposure to adversity and risk for psychopathology have been identified, we made full use of available ALSPAC data and coded each sensitive period model based on the time periods when adversity was measured in the ALSPAC dataset, enabling us to use a more fine-grained set of measures (i.e., specific ages of exposure) to detect possible sensitive periods, an approach found elsewhere to be informative (Dunn *et al.* 2013). However, to facilitate interpretation of our findings and compare our results to prior studies, which have used similar but slightly broader age categories to define sensitive

Table 1. Description of the life course theoretical models tested in the current analysis, using exposure to abuse as an example

Life course model tested	Definition	Number of variables	Specific variables entered into the LARS model
Accumulation of risk (by duration)	Sum of the total number of time periods of exposure to a specific adversity. To test whether the total number of time periods of exposure to a given adversity explains the most variability in psychopathology outcomes	1	abuse_accumulation = count of the number of time periods exposed to abuse (range 0–6)
Sensitive period	A single developmental time period at which there can be exposure to adversity. To test if presence v. absence of a given adversity at a specific time period explains the most variability in psychopathology outcomes	6	abuse_period1 = exposed (1) v. unexposed (0) at time period 1 (18 months); abuse_period2 = exposed (1) v. unexposed (0) at time period 2 (30 months); abuse_period3 = exposed (1) v. unexposed (0) at time period 3 (42 months); abuse_period4 = exposed (1) v. unexposed (0) at time period 4 (57 months); abuse_period5 = exposed (1) v. unexposed (0) at time period 5 (69 months); abuse_period6 = exposed (1) v. unexposed (0) at time period 6 (81 months)
Recency	Sum of the total number of time periods of exposure to a given adversity, with each time period weighted by the age in years of the child during the exposure. To test if temporal proximity to adversity events explains the most variability in psychopathology outcomes	1	abuse_recency = abuse_period1 exposed (1) v. unexposed (0)*(18/12) + abuse_period2 exposed (1) v. unexposed (0) *(30/12) + abuse_period3 exposed (1) v. unexposed (0) *(42/12) + abuse_period4 exposed (1) v. unexposed (0) *(57/12) + abuse_period5 exposed (1) v. unexposed (0) *(69/12) + abuse_period6 exposed (1) v. unexposed (0) *(81/12)

For each type of adversity, we generated three sets of encoded variables: (a) a single variable denoting the total number of time periods of exposure to a given adversity, to test the accumulation hypothesis (coded as 0–6); (b) a set of variables indicating presence v. absence of the adversity at a specific developmental stage, to test the sensitive period hypothesis; and (c) a single variable denoting the total number of time periods of exposure, with each exposure linearly weighted by age (in months) of the child during the measurement time period, to test the recency hypothesis; this variable assumed a linear increase in the effect of exposure over time and weighted more recent exposures more heavily than distally-occurring ones, allowing us to determine whether more recent exposures were more impactful (Smith *et al.* 2016). This weighted recency variable is distinguished from the last sensitive period model, which captures only the most recent exposure.

periods (Kaplow & Widom, 2007; Andersen *et al.* 2008; Slopen *et al.* 2014; Dunn *et al.* 2016), we present our results (examining each specific age stage of exposure) according to three developmental periods – *very early childhood*, ages 0–3; *early childhood*, ages 4–5; *middle-childhood*, ages 6–7.

Child psychopathology

Child emotional and behavioral problems were assessed using the Strengths and Difficulties Questionnaire (SDQ) (Goodman, 1997; Goodman, 2001), which mothers completed by mail when the child was 8 years old. The SDQ is one of the most commonly used dimensional ratings of child psychopathology in epidemiology studies and has excellent psychometric properties (Muris *et al.* 2003; Ezpeleta *et al.* 2013). The SDQ contains 25 items, rated on a three-point scale (0 = not true, 1 = somewhat true, or 2 = certainly true), capturing the child's behavior and feelings within the past 6 months. We calculated a total SDQ score by summing across items on the first four subscales (conduct problems; emotional symptoms; hyperactivity; peer problems; range 0–40), with higher scores indicating more emotional and behavioral difficulties ($\alpha = 0.82$). This total score has been shown in studies from across the globe to correlate highly with questionnaire and interview measures of psychopathology, including the Child Behavior Checklist as well as clinician-rated diagnoses of child mental disorder (Goodman *et al.* 2010; Goodman & Goodman, 2011).

Covariates

We controlled for the following covariates, measured at child birth: child race/ethnicity; pregnancy size; number of previous pregnancies; maternal age; maternal marital status; homeownership; highest level of maternal education; and parent social class (see Supplementary Materials for coding). We also controlled for levels of maternal psychopathology symptoms measured during pregnancy with the Edinburgh Postnatal Depression Scale (Cox *et al.* 1987) to reduce potential impacts of both confounding and common rater bias (Podsakoff *et al.* 2003), as mothers reported about their child's emotional and behavioral problems, mothers were the primary reporters of their child's exposure to adversity, and maternal mood or other factors may influence reports of adversity exposure (Holt *et al.* 2008) and psychopathology (Chilcoat & Breslau, 1997; Ringoot *et al.* 2015). The covariates were included because they were found in our study to be potential confounders or were routinely included in birth cohort studies of child health outcomes (Hibbeln *et al.* 2007; Suren *et al.* 2014). Both sets of results with and without adjustment for maternal psychopathology symptoms are presented to facilitate future replication efforts.

Analyses

After conducting univariate and bivariate analyses to examine the distribution of covariates and exposure to adversity in the total analytic sample, we compared the theoretical models using a two-stage structured life course modeling approach (SLCMA) originally developed by Mishra (Mishra *et al.* 2009) for analyzing repeated, binary exposure data across the life course. Relative to a more traditional regression model, the main advantage of the SLCMA is that it provides a structured and unbiased way to compare multiple competing theoretical models simultaneously and identify the most parsimonious explanation for the observed outcome variation.

In the first stage, we followed the approach of Smith (Smith *et al.* 2015) and entered the set of variables described previously into a Least Angle Regression (LARS) procedure (Efron *et al.* 2004) in order to identify, separately for each type of adversity, the single theoretical model (or potentially more than one model working in combination) that explained the most variability in child emotional and behavioral problems. Thus, four separate LARS models were conducted, corresponding to each type of adversity, separately for boys and girls. We used a covariance test (Lockhart *et al.* 2014) and examined elbow plots (Fig. 1) to determine whether the selected models were supported by the ALSPAC data. Compared with other variable selection procedures, including stepwise regression, the SLCMA has been shown to not over-inflate effect size estimates (Efron *et al.* 2004) or bias hypothesis tests (Lockhart *et al.* 2014). Compared with other methods for the structured approach, LARS has been shown to have greater statistical power and not bias subsequent stages of analysis (Smith *et al.* 2015). Notably, the covariance test *p* values derived from the LARS also account for the other variables being (sequentially) tested in the procedure, making the type I error rate controlled for each type of adversity.

All analyses were stratified by sex. To adjust for potential confounding, we regressed each encoded variable on the covariates and implemented LARS on the regression residuals (Smith *et al.* 2016).

In the second stage, the theoretical models determined by a covariance test *p* value threshold of 0.05 in the first stage (which appeared before the elbow; see Fig. 1) were carried forward to a multiple regression framework, where measures of effect were estimated for all selected hypotheses. The goal of this second stage was to determine the contribution of a selected theoretical model after adjustment for covariates as well as other selected theoretical models, in instances where more than one theoretical model was chosen in the first stage. To reduce potential bias and minimize loss of power due to attrition, we performed multiple imputation in both stages (see Supplementary Materials).
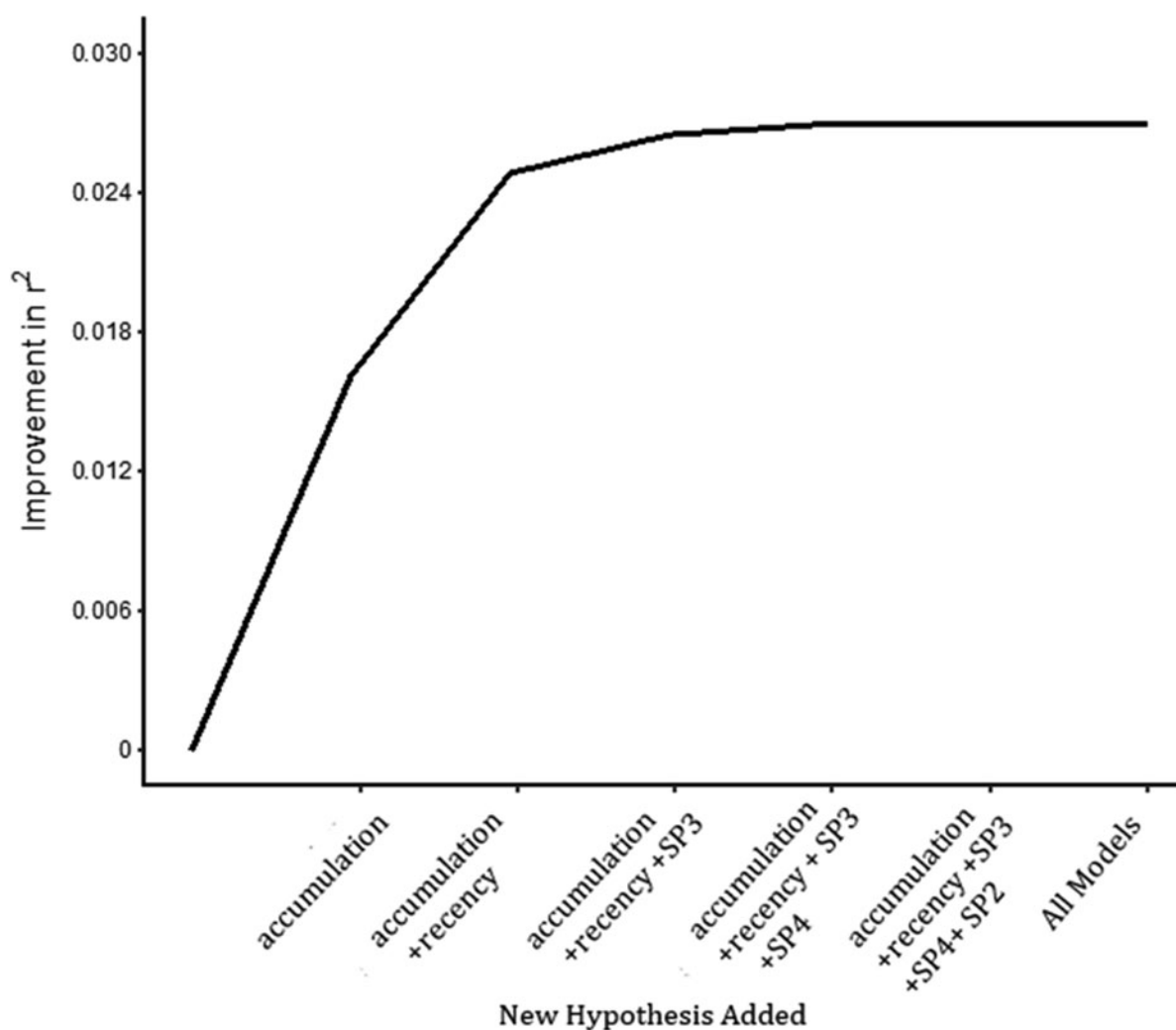

Fig. 1. Elbow plot illustrating the LARS variable selection procedure testing life course models. LARS begins by first identifying the single variable with the strongest association with the outcome; it then identifies the combination of two variables with the strongest association, followed by three variables, and so on, until all variables are included. LARS therefore achieves parsimony by identifying the smallest combination of encoded variables that explain the most amount of outcome variation. In addition to a covariance test, which is calculated at each stage of the LARS procedure and tests the null hypothesis that adding the next encoded variable does not improve r^2, results can also be summarized in an 'elbow plot,' showing the increase in overall model r^2 as additional predictors were added to the model. The point where this plot levels off indicates the point of diminishing marginal improvement to the model goodness-of-fit from adding additional predictors, suggesting that the predictors included in the model at this point represent an optimal balance of parsimony and thoroughness. In this example, both accumulation and recency were selected in the best fitting models. SP, sensitive period.

Results

Sample characteristics and distribution of exposure to adversity

The imputed analytic sample ($n = 7476$) was gender-balanced (49.2% girls) and comprised of predominately White (94.6%) children from families whose parents were married and owned their home (Supplementary Table 1).

Approximately half of the children in this analytic sample (49%; $n = 3694$) experienced at least one adversity. As shown in Table 2, the most commonly experienced adversity, for both boys and girls, was financial stress (32% girls; 30% boys). Parent legal problems were the least reported (6% in girls and boys).

Age at exposure to adversity somewhat varied by type. For instance, caregiver physical or emotional abuse was more common in middle childhood than infancy (Table 2). However, the remaining adversities were primarily reported with the same frequency across time.

Within each adversity type, exposure was correlated over time (Table 3; average correlations: caregiver abuse $r = 0.61$; abuse by anyone $r = 0.44$; legal problems $r = 0.52$; financial stress $r = 0.54$). In general, neighboring time points were more highly correlated than distant time points.

However, across adversity types, the exposures were only modestly correlated (Supplementary Table 3 and Supplementary Fig. 1; average correlation across adversity types $r = 0.24$).

Both exposure to any adversity and child emotional and behavioral problems were patterned by socio-demographic factors, including sex, and socioeconomic status (Supplementary Table 1).

Model selection

Table 4 shows the models selected by the LARS procedure for each adversity type, in boys and girls. Overall, recency was the theoretical model best supported by the data for the abuse-related adversities.

In girls, recency of caregiver physical or emotional abuse explained 1.55% of the variation in child emotional and behavior problems. The combination of recency and exposure to physical or sexual abuse during middle childhood (at 6.75 years of age), the last time point of assessment for this exposure, explained 2.35% of the variation in child emotional and behavior problems. Further, both accumulation and exposure to financial stress during infancy (at 8 months of age) were selected. Accumulation of parent legal problems explained 0.51% of the variation in emotional and behavioral problems.

In boys, accumulation was the best theoretical model chosen for caregiver physical or emotional abuse, explaining 1.71% of the variation in child emotional behavior problems. Recency of sexual or physical abuse explained 1.68% of the variation in boys. Moreover, accumulation was selected as the single best fitting model for financial stress ($r^2 = 1.39\%$), whereas for parent legal problems, exposure during infancy (at 8 months of age) was most important ($r^2 = 0.29\%$).

Model selection results were similar after adjusting for maternal depression (Supplementary Table 4), though two differences are noted. In girls, exposure to financial stress during the sensitive period 1 (at 8 months of age) was not significantly associated with emotional and behavior problems; accumulation was the theoretical model best supported by the data ($r^2 = 0.76\%$). In boys, recency of caregiver physical or emotional abuse replaced accumulation as the best-supported model ($r^2 = 0.89\%$).

Out of all combinations of theoretical models considered for all adversities, financial stress was the type of adversity that explained the largest amount of variation in psychopathology symptoms among girls ($r^2 = 3.08\%$). Among boys, sexual or physical abuse was the most strongly associated type ($r^2 = 1.68\%$).

Effect estimation

After identifying the theoretical models shown in the first stage to explain the most outcome variation, we then entered these theoretical models into a multiple linear regression. As shown in Fig. 2, Panel A, which presents these results for girls, we found that girls exposed to caregiver physical or emotional abuse during more recent developmental periods had the largest increase in emotional and behavior problems as compared to those exposed during earlier time periods (an increase of 0.27 for every additional year of exposure, 95% CI = 0.22, 0.32; standardized mean difference for exposure at age 6 ($SMD_{6y} = 0.34$)). The association with exposure to sexual or physical abuse increased linearly with the proximity of exposure, such that girls exposed at more recent developmental periods had the most emotional and behavioral problems ($\beta = 0.24$; 95% CI = 0.14, 0.35; $SMD_{6.75y} = 0.35$). Further, the association with exposure to sexual or physical abuse also increased with proximity of exposure, but in a nonlinear fashion such that exposure at age 6.75 was a particularly sensitive period, conferring an additional increase in symptoms ($\beta = 1.78$; 95% CI = 0.21, 3.35; $SMD_{6.75y} = 0.38$). For financial stress, where two theoretical models were chosen, each time period of exposure was linearly associated with an increase of 0.54 (95% CI = 0.35, 0.73; $SMD_5 = 0.58$), though girls exposed to financial stress at age 8 months had an additional increase of 1.05 (95% CI = 0.41, 1.68; $SMD_{8mo} = 0.22$) in the measure of emotional and behavior problems. More time periods of exposure to parent legal problems were also linearly associated with increasing emotional and behavior problems (an increase of 0.82 per event, 95% CI = 0.45, 1.19; $SMD_6 = 1.04$).

As shown in Fig. 2, Panel B, boys exposed to sexual or physical abuse more recently had higher emotional and behavioral problems (an increase of 0.28 per additional year of exposure; 95% CI = 0.21, 0.34; $SMD_{6.75y} = 0.36$). More time periods of exposure to either caregiver physical or emotional abuse or financial stress were linearly associated with increasing emotional and behavior problems (increases of 1.04 per event, 95% CI = 0.82, 1.26; $SMD_6 = 1.18$, and 0.70 per event, 95% CI = 0.53, 0.86; $SMD_5 = 0.66$ respectively). Exposure to parent legal problems at 8 months of age was associated with increased child psychopathology ($\beta = 3.03$; 95% CI = 1.43, 4.64; $SMD_{8mo} = 0.57$).

Discussion

The primary finding of this study is that child psychopathology symptoms were largely explained by the accumulation and recency of exposure to adversity, rather than sensitive periods. Specifically, for either type of abuse, we found that more recently occurring exposures were generally more harmful, as the LARS procedure most frequently selected the recency model for these types of adversity. This finding is consistent with at least one prior study testing the recency hypothesis (Shanahan *et al.* 2011) and other work showing that the depressogenic effects of adversity are elevated in the same month or month after the event (Kendler *et al.* 1999) or the same year of exposure (Dunn *et al.* 2012). Accumulation was the second theoretical model

Table 2. Exposure to childhood adversity overall and by age at exposure

	Abuse								Stress							
	Caregiver physical or emotional abuse				Sexual or physical abuse (by anyone)				Financial stress				Parent legal problems			
	Female		Male		Female		Male		Female		Male		Female		Male	
	N	(%)	N	(%)	N	(%)	N	(%)	N	(%)	N	(%)	N	(%)	N	(%)
Unexposed	2159	83	2273	83	2335	90	2352	86	1982	68	2098	70	2402	94	2531	94
Exposed	446	17	471	17	270	10	394	14	936	32	890	30	160	6	155	6
Timing of exposure																
Very early childhood																
Age 8 mo.	92	3.5	103	3.8	–	–	–	–	351	12	334	11	25	1	32	1.2
Age 1.5/1.75	103	4	100	3.7	49	1.9	70	2.6	350	12	346	12	37	1.5	29	1
Age 2.5/ 2.75	138	5.3	168	6.1	75	2.9	123	4.5	344	12	328	11	44	1.7	47	1.8
Early childhood																
Age 3.5	–	–	–	–	59	2.3	86	3.1	375	13	338	11	–	–	–	–
Age 4/4.75	147	5.6	125	4.6	56	2.2	117	4.3	–	–	–	–	44	1.7	37	1.4
Age 5/5.75	166	6.4	197	7.2	60	2.3	101	3.7	–	–	–	–	33	1.2	36	1.3
Middle childhood																
Age 6/6.75	168	6.5	139	5.1	61	2.3	94	3.4	–	–	–	–	34	1.3	36	1.3
Age 7	–	–	–	–	–	–	–	–	304	10	297	9.6	–	–	–	–

Percentages for each age represent proportions of those exposed out of the total population.
– indicates that the variable was not assessed at the corresponding time point.

Table 3. Tetrachoric correlations between childhood adversities

	Caregiver physical or emotional abuse (N = 5349)						Sexual or physical abuse (by anyone) (N = 5351)					
Age	8 mo	1.75	2.75	4	5	6	*Age*	1.5	2.5	3.5	4.75	5.75	6.75
8 mo	1	–	–	–	–	–	1.5	1	–	–	–	–	–
1.75	0.71	1	–	–	–	–	2.5	0.5	1	–	–	–	–
2.75	0.61	0.7	1	–	–	–	3.5	0.41	0.44	1	–	–	–
4	0.5	0.6	0.69	1	–	–	4.75	0.39	0.43	0.5	1	–	–
5	0.51	0.55	0.61	0.65	1	–	5.75	0.35	0.38	0.42	0.48	1	–
6	0.49	0.54	0.56	0.65	0.7	1	6.75	0.32	0.46	0.38	0.48	0.64	1
	Financial stress (N = 5906)						Parent legal problems (N = 5248)					
Age	8 mo	1.75	2.75	5	7		*Age*	8 mo	1.75	2.75	4	5	6
8 mo	1	–	–	–	–		8 mo	1	–	–	–	–	–
1.75	0.69	1	–	–	–		1.75	0.64	1	–	–	–	–
2.75	0.66	0.74	1	–	–		2.75	0.58	0.65	1	–	–	–
5	0.48	0.49	0.55	1	–		4	0.48	0.55	0.72	1	–	–
7	0.39	0.45	0.43	0.56	1		5	0.36	0.38	0.48	0.61	1	–
							6	0.4	0.45	0.39	0.5	0.58	1

selected most frequently. Dozens of studies have shown that chronic or cumulative exposure to adversity is harmful to mental health and other outcomes (Evans *et al.* 2013).

However, only two clear sensitive periods were identified. The first was for financial stress in girls, where we found that both accumulation and exposure during very early childhood were most strongly associated with child emotional and behavior problems. That is, while each additional time-period of exposure was linearly associated with an increase in psychopathology symptoms, girls first exposed to financial stress at age 8 months had even worse emotional and behavioral problems with more accumulated exposure. The second sensitive period observed was for parent legal problems, where we found that exposure at age 8 months had the strongest association with psychopathology symptoms. Therefore, our results on this occasion provide limited additional insight compared with studies that only examined whether or not a child was exposed.

Why did so few sensitive periods emerge? Our inability to identify sensitive periods was surprising, given that numerous animal studies have found time-dependent effects of adversity on a range of outcomes, including not only anxious/depressive symptoms (Raineki *et al.* 2012), but also social, emotional, and behavioral processes (e.g., fear conditioning, stress reactivity, aggressive behavior, Sanchez *et al.* 2001; Holmes *et al.* 2005; Veenema, 2009), and brain structure and function (Liu *et al.* 2012; Makinodan *et al.* 2012). However, in the human literature, there is mixed support for the existence of sensitive periods shaping risk for psychopathology. Research on the importance of the developmental timing of child maltreatment on depression risk provides a good illustration of such inconsistencies. Several prospective studies have found higher levels of internalizing symptoms in early childhood (Keiley *et al.* 2001) and depressive symptoms in early (Thornberry *et al.* 2010) and early to mid-adulthood (Kaplow & Widom, 2007) among individuals exposed

Table 4. Results of LARS models on multiply imputed data, stratified by sex

	Female (N = 3676)		Male (N = 3800)	
	Model(s) selected	r^2 explained (%)	Model(s) selected	r^2 explained (%)
Abuse				
Caregiver physical or emotional abuse	Recency	1.55	Accumulation	1.71
Sexual or physical abuse	Recency and sensitive period in middle childhood (age 6.75 years)	2.35	Recency	1.68
Stress				
Financial Stress	Accumulation and sensitive period in very early childhood (age 8 months)	3.08	Accumulation	1.39
Parent legal problems	Accumulation	0.51	Sensitive period in very early childhood (age 8 months)	0.29

The table indicates the set of theoretical models chosen by the LARS, after adjusting for covariates.
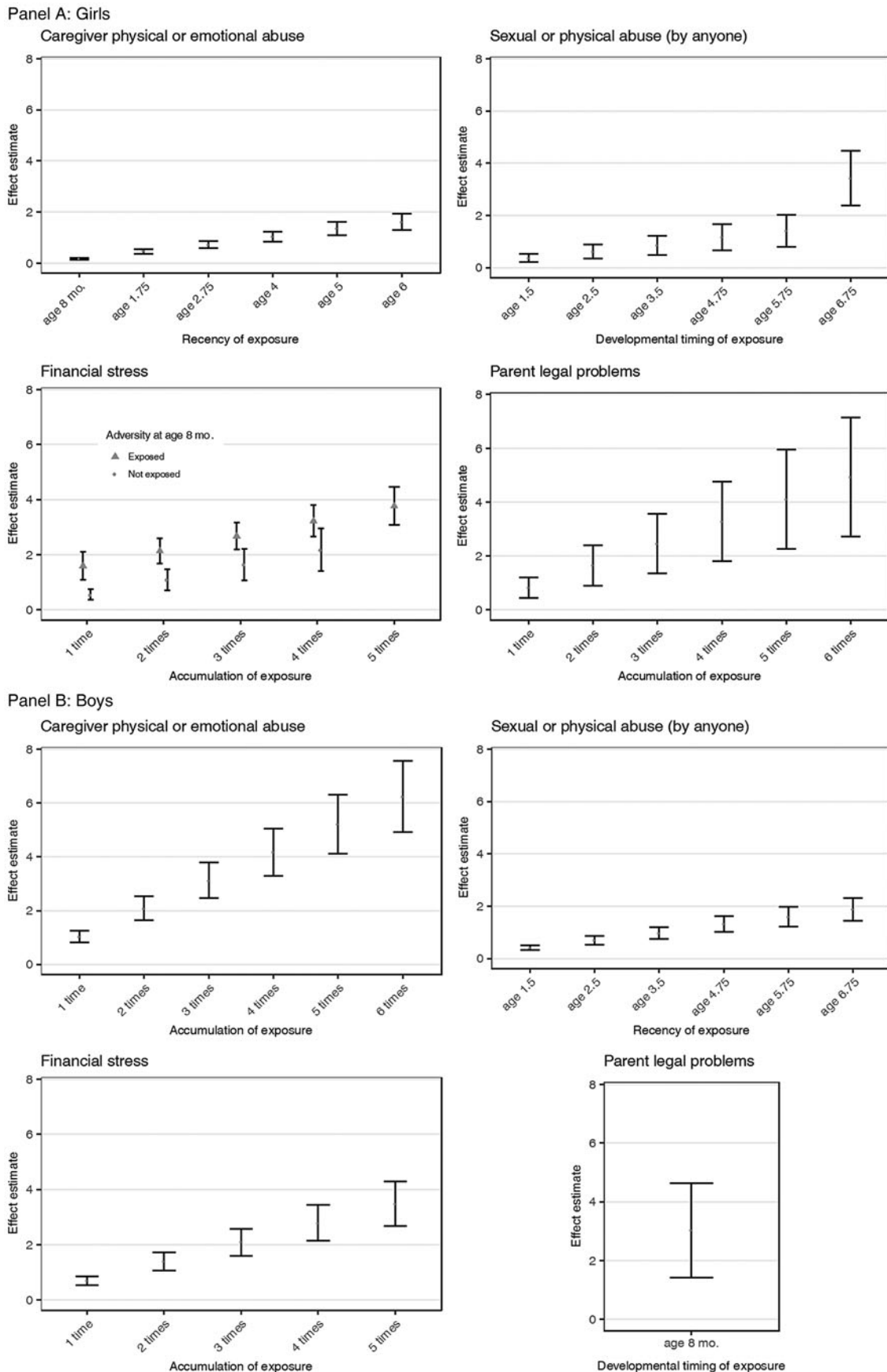
Panel A: Girls

Fig. 2. Effect estimates for exposure to adversity, stratified by sex. The effect estimates and corresponding 95% confidence intervals demonstrate the increase in SDQ scores for exposure to adversity during a certain time point or number of exposures. For accumulation, an increase in the number of time points exposed corresponds to a greater increase in SDQ score. For recency, exposure to adversity during later time points corresponds to a greater increase in SDQ score.

to child maltreatment before age 5 compared to those who were either never exposed or exposed during later stages. However, several prospective studies have found no effect of maltreatment timing (Manly *et al.* 2001; English *et al.* 2005; Jaffee & Maikovich-Fong, 2011) or that maltreatment exposure during adolescence is more harmful than exposure during earlier developmental stages (Thornberry *et al.* 2001; Harpur *et al.* 2015). Conflicting findings could reflect differences in the length of time between the onset of adversity and measurement of the outcome, showcasing more 'recency' rather than sensitive period effects. Our future research will perform similar analyses in relation to outcomes measured during adolescence and adulthood, which will help evaluate the longer-term effects of adversity on both the onset and course of psychopathology symptoms and help determine whether the lack of distinct sensitive periods within childhood is common to other outcomes. Heterogeneity in the literature could be explained by the fact that there are unlikely sensitive periods *for* psychopathology *per se*. Instead, adversity likely disrupts multiple intermediate processes linked to psychopathology, including attention and emotion recognition; each of these domains could have their own sensitive period.

As expected, sex differences were observed. For example, there were instances when more than one theoretical model was operating simultaneously to produce mental health outcomes in one sex, but a single theoretical model was operating for another. Importantly, these differences did not appear driven by sex differences in the prevalence of exposure, as boys and girls were exposed to each of these adversities at the same frequency. Future studies are needed to understand the factors giving rise to these sex differences and replicate findings regarding the importance of developmental timing, as few studies in this area have been conducted (Najman *et al.* 2010*a*, *b*).

Although the variance explained by each of these best-fit models may at first appear small, it bears noting that these life course models are examining a single adversity type in a large population-based sample, as opposed to a cumulative adversity score in a clinical sample. Moreover, unlike models examining the variance explained by a given exposure, the examined life course models are examining effect sizes for the *temporal patterns* of certain exposures. Thus, the size of the reported R^2 values is on par with what we might expect given the temporal specificity of the models and the population-based nature of the sample.

This study has several strengths. We conducted these analyses in a large, longitudinal, and population-based sample of children, which minimized the likelihood of retrospective recall bias that is common among studies of childhood adversity and allowed us to evaluate the short-term consequences of adversity on psychopathology symptoms. We also applied a novel analytic approach that enabled us to simultaneously compare these theoretical models and evaluate the impact of each theoretical model to each adversity type. Comparisons of these models by type of adversity may contribute to insights about the mechanisms underlying psychopathology risk. Information about the types and features of adversity that are most strongly associated with childhood psychopathology may also help identify highest priority points for intervention.

We also considered exposures individually, rather than simultaneously, which was arguably both a strength and limitation. On the one hand, focusing on one type of childhood adversity without accounting for the impact of highly correlated exposures can artificially inflate effect estimates for the single adversity type (Dong *et al.* 2005; Green *et al.* 2010). However, in our sample, the adversities examined were only modestly correlated with each other. Of note, the clustering of different types of adversity experiences with typically high co-occurrence, such as physical and emotional abuse and sexual and emotional abuse (which was done through the combined format in the questionnaire), may help account for our lower inter-correlations between adversity experiences in this sample.

On the other hand, attention to specific adversity types – and in particular the time-course of exposure to these adversity types – proved meaningful for understanding adversity-specific associations to risk for psychopathology. The finding that different life course models differentially explained the association between childhood adversity and psychopathology symptoms suggests that grouping adversity experiences could have obscured these distinctions. An important next step would be to consider ways to examine multiple adversities simultaneously so that meaningful information could be gleaned without simply summing across the number of adversities experienced (McLaughlin & Sheridan, 2016).

Several limitations are noted. The use of single items to capture adversity could affect the precision of these estimates. However, the prevalence of these adversities, including those capturing experiences of abuse, were comparable with estimates derived from nationally-representative samples (Gilbert *et al.* 2009; McLaughlin *et al.* 2012). As with any longitudinal study, there was attrition over time, which we attempted to address using multiple imputation. We were also unable to examine the impact of experiencing multiple adversities simultaneously because these adversities were measured at different time points. Furthermore, most of the socio-demographic covariates were only measured at birth, which may be problematic as some of these variables could be time-varying, including indicators of socioeconomic status. Finally, although we controlled for several potential confounding factors, including maternal psychopathology, it is possible that residual confounding may remain, including through unmeasured genetic factors that shape both exposure and outcome (i.e., gene–environment correlation). As more genetic variants associated with neuropsychiatric phenotypes emerge from genome-wide association studies, future studies will be better positioned to ensure that study results are not explained by genetic factors.

In summary, our results suggest that no single theoretical model best captures the relationship between adversity and mental health problems, but rather than depending on sex and the type of exposure, adversities can operate through different pathways. These findings underscore the importance of measuring the characteristics of adversity, which can help further elucidate the most important environmental risk factors shaping child mental health.

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Supplemental Materials

Measures

Child emotional and behavioral problems were assessed using total scores derived from the Strengths and Difficulties Questionnaire (SDQ) (Goodman, 1997, Goodman, 2001). This total score had a greater internal consistency reliability coefficient (α=0.82) relative to each individual subscale (Emotional Symptoms Scale α=0.69; Conduct Problem Scales α=0.59; Hyperactivity Scale α=0.80; Peer Problems Scale =0.61).

We controlled for the following covariates, measured at child birth: *child race/ethnicity* (0=non-White; 1=White); *pregnancy size* (0=single; 1=multiple); *number of previous pregnancies* (between 0-3+); *maternal marital status* (0=never married; 1=widowed/divorced/separated; 2=married); *highest level of maternal education* (1=less than O-level, 2=O-level, 3=A-level, 4=Degree or above); *maternal age* (0=ages 15-19, 1=ages 20-35, 2=age>35); *homeownership* (0=mortgage/own home; 1=rent home; 2=other); and *parent social class* (i.e. the highest social class of either parent: 1=foreman; 2=manager; 3=supervisor; 4=lending hand; 5=self-employed; 6=none of these/missing). Notably, these latter two indicators of socioeconomic status were measured infrequently and thus were difficult to examine after birth.

Missingness

In the current study, the analytic sample consisted of children that had complete outcome data measured at age 8. Children that had complete data (n=4350) differed from children who were missing on any covariate or exposure (n=3126) with respect to most covariates, including race and socioeconomic status (**Supplementary Table 2**). Additionally, children missing any data were more likely to have been exposed to adversity at any time (60.21% versus 43.24%; p<0.0001) and had higher scores on the measure of child emotional and behavioral problems, indicating more problems (mean +/- SD: 8.16 +/- 5.29 versus 7.56 +/- 5.07; p<0.0001).

Multiple Imputation and Statistical Analyses

To reduce potential bias and minimize loss of power due to attrition, we performed multiple imputation, separately for each exposure, using logistic regression in 20 datasets with 25 iterations each among all children with complete data on the outcome. Variables were included in the imputation models following the guidance of van Buuren and colleagues (van Buuren et al., 1999, van Buuren and Groothuis-Oudshoorn, 2011) as well as prior research with imputation in the ALSPAC dataset (Ramchandani et al., 2008, Evans et al., 2012). The following variables were allowed to enter the imputation models: all covariates (including maternal depression), later exposures to the same adversity (if available, measured through age 9), exposure to the other adversities, later outcomes (behavior symptoms and internalizing symptoms measured at ages 10, 11, 13, 16, and 18), and other maternal behavior measures (i.e., alcohol intake and smoking behavior). Variables uncorrelated with the missing variable (r<0.10) were excluded from the imputation model (van Buuren et al., 1999, van Buuren and Groothuis-Oudshoorn, 2011). Imputation was performed with chained equations (Azur et al., 2011) with the *mice* package in R (van Buuren and Groothuis-Oudshoorn, 2011). To reduce noise in estimation of effect estimates, we did not impute the outcome (White et al., 2011). For each adversity, we assessed the convergence of the imputation model and the distribution of imputed data as compared to the observed data.

After imputation, there were 7,476 children in the analytic sample. We then achieved a single dataset for analysis by implementing LARS on the covariance structure among all variables, estimated by averaging the covariance structure across all multiply imputed datasets. This allowed us to avoid potential problems arising from different model selections across multiply imputed datasets (Wood et al., 2008).

After selecting the best fitting models from Stage 1, we performed a linear regression of the theoretical model chosen on each of the 20 multiply imputed datasets and pooled effect estimates (regression coefficients) across datasets using Rubin's rules (van Buuren and Groothuis-Oudshoorn, 2011, Rubin, 1987). We used the p-value from the covariance test to calculate unbiased confidence intervals for the effect estimates.(Smith et al., 2015, Lockhart et al., 2014)

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Supplemental Table 1. Distribution of covariates in the total sample and by exposure to any childhood adversity and by levels of child emotional and behavioral problems (N=7476)

	Total Sample		Exposure to any adversity				Child Emotional and Behavioral Problems			
	%	N	%	N	χ^2	*p*-value	Mean	SD	*F*	*p*-value
Gender					2.75	0.0970			92.48	< 0.0001
Males	50.83	3800	50.37	1914			8.34	5.41		
Females	49.17	3676	48.42	1780			7.20	4.83		
Race					6.90	0.0086			0.05	0.8263
White	96.49	6704	49.84	3341			7.67	5.10		
Non-White	3.51	244	58.60	143			7.75	5.37		
Maternal Education					64.10	< 0.0001			19.07	< 0.0001
less than O-level	21.51	1518	58.17	883			8.40	5.58		
O-level	35.32	2493	50.42	1257			7.81	5.03		
A-level	26.58	1876	48.99	919			7.25	4.88		
Degree or Above	16.59	1171	42.95	503			7.17	4.87		
Pregnancy Size					3.18	0.0743			0.18	0.6747
Single	97.73	7306	49.25	3598			7.77	5.17		
Multiple (2+)	2.27	170	56.47	96			7.94	4.76		
Maternal Marital Status					138.32	< 0.0001			24.53	< 0.0001
Never Married	13.47	957	65.10	623			8.58	5.35		
Widowed/Divorced/Separated	4.98	354	66.10	234			8.59	5.13		
Married	81.54	5792	47.39	2745			7.49	5.04		
Home Ownership					215.12	< 0.0001			37.54	< 0.0001
Mortgage/own home	82.49	5821	46.78	2723			7.46	4.99		
Rent home	14.62	1032	69.86	721			8.95	5.65		
Other	2.89	204	69.12	141			8.01	5.05		
Age of Mother at child birth					24.48	< 0.0001			16.05	< 0.0001
Ages 15-19	1.95	141	71.63	101			10.04	5.89		
Ages 20-35	89.63	6489	50.61	3284			7.72	5.14		
Age >35	8.43	610	50.49	308			7.35	4.64		
Parental Social Class					56.09	< 0.0001			18.44	< 0.0001
Foreman	13.83	1034	41.49	429			6.84	4.70		

	%	N	%	N			Mean	SD		
Manager	36.50	2729	47.45	2143			7.47	4.96		
Supervisor	20.84	1558	51.93	809			7.90	5.18		
Lending Hand	5.31	397	57.43	228			8.42	5.40		
Self-Employed	1.54	115	60.00	69			8.57	4.83		
None of these	21.98	1643	52.59	864			8.56	5.56		
Number of previous pregnancies					51.10	< 0.0001			7.02	0.0001
0	46.49	3269	47.35	1548			8.00	5.14		
1	35.81	2518	51.43	1295			7.38	4.93		
2	13.36	939	56.44	530			7.67	5.38		
3+	4.34	305	64.59	197			7.52	5.73		

	Mean	SD	Mean	SD	t	p-value			Pearson's r	p-value
Maternal depression	5.21	4.57	6.27	4.97	-19.86	< 0.0001			0.258	< 0.0001

Supplemental Table 2. Distribution of covariates, exposure, and outcome, stratified by missingness

	Complete cases		Participants with any missing data			
	%	N	%	N	X²	p-value
Gender					0.55	0.4546
Males	51.17	2434	50.24	1366		
Females	48.83	2323	49.76	1353		
Race					10.92	0.0010
White	96.99	4614	95.39	2090		
Non-White	3.01	143	4.61	101		
Maternal Education					119.60	< 0.0001
less than O-level	18.50	880	27.73	638		
O-level	34.66	1649	36.68	844		
A-level	27.96	1330	23.73	546		
Degree or Above	18.88	898	11.86	273		
Pregnancy Size					12.22	0.0005
Single	98.19	4671	96.91	2635		
Multiple (2+)	1.81	86	3.09	84		
Maternal Marital Status					121.58	< 0.0001
Never Married	10.66	507	19.18	450		
Widowed/Divorced/ Separated	4.27	203	6.44	151		
Married	85.07	4047	74.38	1745		
Home Ownership					147.62	< 0.0001
Mortgage/own home	86.29	4105	74.61	1716		
Rent home	11.31	538	21.48	494		
Other	2.40	114	3.91	90		
Age of Mother at child birth					94.97	< 0.0001
Ages 15-19	0.88	42	3.99	99		
Ages 20-35	89.76	4270	89.37	2219		
Age >35	9.35	445	6.65	165		
Parental Social Class					554.29	< 0.0001
Foreman	16.50	785	9.16	249		
Manager	41.31	1965	28.10	764		
Supervisor	21.69	1032	19.34	526		
Lending Hand	5.15	245	5.59	152		
Self-Employed	1.56	74	1.51	41		
None of these	13.79	656	36.30	987		
Number of previous pregnancies					3.40	0.3335
0	46.71	2222	46.04	1047		
1	36.16	1720	35.09	798		
2	12.99	618	14.12	321		
3+	4.14	197	4.75	108		
Exposure to any adversity					198.51	< 0.0001

No	56.76	2700	39.79	1082		
Yes	43.24	2057	60.21	1637		
	Mean	SD	Mean	SD	*t*	*p*-value
Maternal depression	5.11	4.42	5.66	4.87	-5.39	< 0.0001
Strength and Difficulties Questionnaire	7.56	5.07	8.16	5.29	-4.82	< 0.0001

Supplemental Table 3. Tetrachoric correlations between lifetime adversity exposures (ever exposed vs. never exposed)

Adversity	1.	2.	3.	4.
1. Caregiver physical or emotional abuse	1	---	---	---
2. Sexual or physical abuse by anyone	0.30	1	---	---
3. Financial stress	0.25	0.15	1	---
4. Parent legal problems	0.27	0.21	0.30	1

Supplemental Figure 1. Graphical depiction of tetrachoric correlations between adversity exposures and covariates
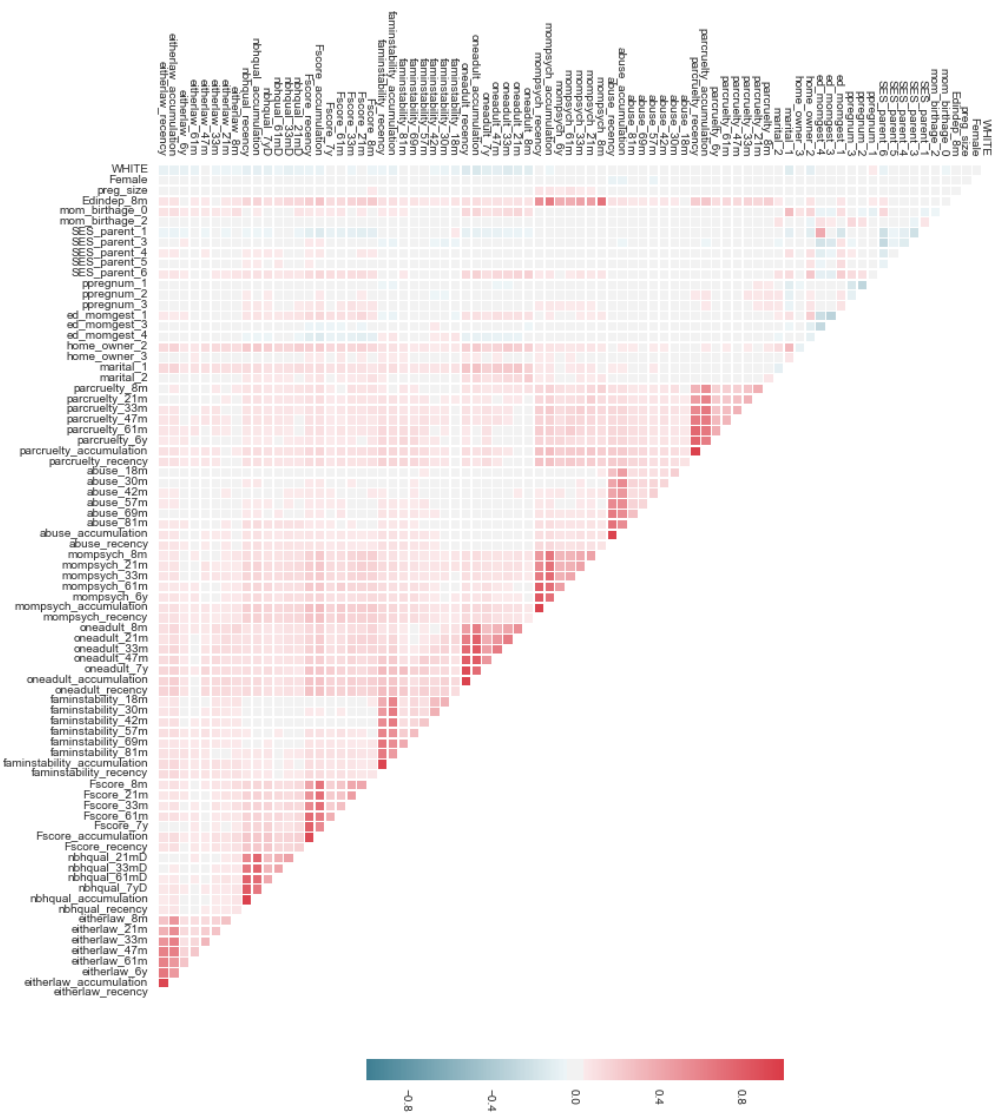
The heat map indicates the strength of the correlations between adversity exposures at each time point and each level of the covariates, with stronger positive correlations denoted in dark red, and stronger negative correlations denoted in dark blue. As shown, most of the heat map is either gray (indicating a correlation close to 0) or pale red (indicating a low- to moderate-strength correlation below r=0.4). The strongest positive correlations were within an adversity type, meaning between models of adversity exposure and measures of that same adversity across time. The weak correlations between socioeconomic status indicators—such as maternal education ("ed_momgest"), home ownership ("home_owner"), and parental social class ("SES_parent")—and financial stress ("Fscore") may in part represent greater social security experienced by British citizens.

Supplemental Table 4. Results of LASSO models on multiply imputed data, adjusted for maternal depression, stratified by sex

	Female (N=3676)		Male (N=3800)	
	Model(s) selected	r₂ explained	Model(s) selected	r₂ explained
Abuse				
Caregiver physical or emotional abuse	Recency	1.63%	Recency	0.89%
Sexual or physical abuse	Recency and Sensitive Period 6 (middle childhood)	1.85%	Recency	1.08%
Stress				
Financial Stress	Accumulation	0.76%	Accumulation	0.55%
Parent legal problems	Accumulation	0.18%	Sensitive Period 1 (very early childhood)	0.21%

The table indicates the set of theoretical models chosen by the LASSO, after adjusting for covariates.
Sensitive Period 1 (infancy) for financial stress refers to 8 months of age.
Sensitive Period 1 (very early childhood) for parent legal problems refers to age 8 months.