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### **Original Article**

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# Associations of offspring birthweight and placental weight with subsequent parental coronary heart disease: survival regression using the walker cohort

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#### **Abstract**

Low birth weight (BW) is consistently correlated with increased parental risk of subsequent cardiovascular disease, but the links with offspring placental weight (PW) are mostly unexplored. We have investigated the associations between parental coronary heart disease (CHD) and offspring BW and PW using the Walker cohort, a collection of 48,000 birth records from Dundee, Scotland, from the 1950s and 1960s. We linked the medical history of 13,866 mothers and 8,092 fathers to their offspring's records and performed Cox survival analyses modelling maternal and paternal CHD risk by their offspring's BW, PW, and the ratio between both measurements. We identified negative associations between offspring BW and both maternal (hazard ratio [HR]: 0.91, 95% confidence interval [CI]: 0.88-0.95) and paternal (HR: 0.96, 95% CI: 0.93-1.00) CHD risk, the stronger maternal correlation being consistent with previous reports. Offspring PW to BW ratio was positively associated with maternal CHD risk (HR: 1.14, 95% CI: 1.08-1.21), but the associations with paternal CHD were not significant. These analyses provide additional evidence for intergenerational associations between early growth and parental disease, identifying directionally opposed correlations of maternal CHD with offspring BW and PW, and highlight the importance of the placenta as a determinant of early development and adult disease.

#### Introduction

Following the investigation linking early life development and adult cardiometabolic outcomes, 1,2 multiple epidemiological studies have investigated associations between offspring birth weight (BW) and parental mortality. Strong and consistent inverse associations have been found between offspring BW and maternal cardiovascular disease (CVD) mortality<sup>3-10</sup> and CVD risk factors such as blood pressure or carotid intima media thickness. 11,12 Studies including fathers have also reported inverse associations with paternal CVD mortality, 3,7-10 although generally weaker compared to the maternal association, suggesting the potential effect of the intrauterine environment conditioning this relationship. 10 Similar associations were found linking BW to CVD mortality in aunts, uncles, and grandparents, 13,14 supporting the role of inheritance of genetic variants influencing fetal growth and increasing CVD risk.<sup>3,7,11</sup> It is hypothesised that the associations between parental CVD and offspring BW result from a matrix of genetic, epigenetic, intrauterine, and other shared environmental influences. 9,15,16

Despite being a vital organ for fetal development, placental weight (PW) is often unavailable for epidemiological studies tying early growth to adult health. Only a few studies have considered the association between placental characteristics and adult disease development, 17-20 and even fewer studies have investigated their association with parental CVD. Davey-Smith et al. did not find significant associations between PW or PW to BW ratio and maternal CVD mortality.<sup>5</sup> However, using a larger sample, Yeung et al. found positive associations between offspring PW to BW ratio and maternal CVD mortality. <sup>21</sup> Apart from the correlation between placental size and fetal development, <sup>22,23</sup> suggesting similar genetic associations to those seen for BW, <sup>24–26</sup> the role of the placenta in the association between fetal growth restriction and parental health outcomes is not yet understood. A recent causal mediation analysis by Sato and colleagues<sup>27</sup> revealed that while maternal polygenic scores for blood pressure measurements are inversely associated with offspring BW, this effect was greatly mediated by placental weight, further adding to the complexity of the early determination of adult cardiometabolic disease.

The Walker cohort<sup>28</sup> is a collection of birth records from Dundee, Scotland, from 1952 to 1970. Walker includes information for 75% of all births in the area for that timeframe, recording PW measurements and details on the mothers and fathers, making it relevant to study links

between parental cardiometabolic disease and birth outcomes. The inclusion of paternal data in these analyses is vital to attempt to discriminate genetic and intrauterine effects from a phenotypic point of view. Using Walker, we previously investigated the associations between offspring BW and PW and parental type 2 diabetes (T2D) risk,<sup>29</sup> identifying novel links between offspring PW and paternal T2D. We hypothesised that parental CHD incidence would be negatively associated with offspring BW and PW, by effect of the genetic inheritance of variants which might reduce fetal growth while also increasing CHD susceptibility. Survival regression analyses of maternal and paternal CHD risk modelled by their offspring BW and PW were performed to test this hypothesis.

#### **Methods**

### Study population and data sources

All individuals included in the analysis were part of the Walker cohort.<sup>28</sup> Offspring BW, PW, gestational age (GA), and sex were documented directly in the Walker records at the time of birth (1951–1968) by obstetricians. For around 70% of the individuals, GA was inferred from the time between the date of birth (DOB) and the last maternal menstrual period, or from the time between 280 days before the DOB and the recorded estimated delivery date, if last menstrual period records were not available. Parental health information was obtained through data linkage using the NHS Scotland Community Health Index unique identifier. Parental CHD and death information were obtained through the SMR01 (hospital admissions) and the National Records Scotland (death records) datasets. The World Health Organization ICD9 and ICD10 codes and National Health Service OPCS-4 codes used to define CHD for these analyses are included in Supplementary Table 1. The national Community Health Index dataset was used to obtain parental DOB and Health Board specific Scottish Index of Multiple Deprivation (HBSIMD, 2019 v2 release), categorising areas according to their deprivation quintile (five meaning least deprived).

#### Data exclusions

Individuals with GA under 37 weeks or over 42 weeks, BW under 2,500g or over 4,500g, or PW under 200g or over 1,000g were excluded from the analysis, to select only healthy term pregnancies and exclude extreme BW and PW measurements. Only singleton pregnancies and firstborn were included. Offspring BW and PW were sex-stratified and standardised through Z-transformation, to compare their effects more appropriately and account for any variation due to offspring sex. The analyses included all identifiable parents living in the area who had not been admitted to hospitals due to CHD causes prior to January 1st 1981, when SMR01 data started being collected routinely. This date defined the study start point. The dataset was supplemented with additional CHD events from the death registry from 1989, when causes of death started being recorded as ICD codes. The study endpoint was defined as the date of the last CHD event (September 12th 2019). After the exclusion process, the final datasets included 13,866 mothers and 8,092 fathers, 91.52% and 92.58% of the total identifiable Walker parents, respectively. A subset of this dataset was used for the supplemental Fine-Gray survival regression analysis, setting the study start in January 1st 1989 due to the need of ICD-coded causes of death from the death records. This dataset included 12,094 mothers and 6,677 fathers. Survival analyses were performed using all individuals with complete information for the explanatory variables.

#### Statistical analyses

All analyses were conducted using the R statistical software<sup>30</sup> version 3.6.2.

#### Summary statistics

All analyses were performed separately for the maternal and paternal datasets. Welch two sample t-tests were used to determine differences in continuous variables by parental CHD status or between the maternal and paternal datasets, using the test of equal proportions for binary variables. Violin plots comparing these differences were built using only offspring whose mother and father could be identified (n = 7,478).

#### Survival analysis study design

Two sets of survival models were built to analyse parental risk of CHD. Cox survival regression<sup>31</sup> was used to investigate the association between offspring BW, PW, and PW:BW ratio and parental CHD risk, defined by CHD-related hospital admission or death by CHD (collected from 1989). The event time was defined as the period between the start of the study (January 1st 1981) and the first CHD event, or until censoring due to death, loss of follow-up, or no event presented. Sex-stratified Z-transformed offspring BW, PW, BW and PW together (BW + PW), and PW to BW ratio (PW:BW) were used as main explanatory variables in different models within each set. The BW and PW models were built to independently assess their contribution to parental CHD risk. The BW + PW models were built to investigate these variables accounting for each other, particularly to provide an estimate for PW accounting for BW. The PW:BW ratio models were built to assess the association such variable as a measure of placental efficiency. Offspring GA, parental age in 1981, and parental HBSIMD were included as additional explanatory variables. Due to the aged population of the study and the lack of follow-up before 1981, supplemental Fine-Gray survival models<sup>32</sup> were built to analyse parental risk of CHD (defined as hospital admission only) accounting for the competing risk of death from causes other than CHD. January 1st 1989 was set as the start of the study for these analyses, as ICD codes for cause of death were not available before then. The Fine-Gray models also included offspring GA, parental age in 1989, and parental HBSIMD as additional variables. The Cox and Fine-Gray regression analyses were performed using the *survival*<sup>33</sup> and *cmprsk*<sup>34</sup> packages, respectively. The effect sizes of the covariates were reported as hazard ratios (HR) for the Cox models, and subdistribution hazard ratios (SHR) for the Fine-Gray models. Scaled Schoenfeld residuals tests<sup>35</sup> were performed to determine violations of the proportional hazards assumption, supported with observational assessment of Schoenfeld residuals against event time plots, performed using the Cox models. In order to account for violations of the proportionality of hazards, any variables with time-dependent effects were adjusted including an interaction term with a logarithmic function of time. Cumulative incidence curves for parental CHD by age and offspring BW, PW, and PW:BW ratio quartiles were built using the *cmprsk* package.<sup>34</sup> The power calculations for the Cox models were performed using the powerSurvEpi package.36

#### Results

### Parental CHD and death summary statistics

Table 1 shows the characteristics of the parents included in the analyses and their offspring. Overall, 23.9% and 36.7% mothers and fathers, respectively, had developed CHD. Fathers had CHD

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Table 1. Summary statistics for the maternal and paternal datasets

	Maternal Dataset (n = 13,866)	Missing	Paternal Dataset (n = 8092)	Missing	p value
Offspring					
BW (g)	3375.83 ± 412.21	0.11%	3389.45 ± 411.60	0.10%	0.018
PW (g)	651.53 ± 117.91	55.03%	654.15 ± 120.21	48.22%	0.270
PW:BW Ratio	0.19 ± 0.03	55.20%	0.19 ± 0.03	48.27%	0.743
GA (weeks)	39.89 ± 1.29	2.54%	39.89 ± 1.28	2.49%	0.743
Age at Study End (y)	59.91 ± 4.68	0.17%	59.15 ± 4.51	0.14%	<0.001
Sex	F: 6025 (43.45%) M: 7841 (56.55%)	0%	F: 3300 (40.78%) M: 4792 (59.22%)	0%	<0.001
Parent					
CHD Event	3308 (23.86%)	-	2967 (36.67%)	-	<0.001
Age at CHD Event (y)	69.92 ± 10.70	1.51%	67.68 ± 10.41	1.40%	<0.001
Death	8559 (61.73%)	-	6035 (74.58%)	-	<0.001
Age at Death (y)	76.53 ± 10.58	1.51%	75.42 ± 9.79	1.40%	<0.001
HBSIMD	1: 4487 (33.91%)	4.58%	1: 2465 (31.46%)	3.17%	< 0.001
	2: 2929 (22.14%)		2: 1661 (21.20%)		
	3: 1635 (12.36%)		3: 977 (12.47%)		
	4: 1967 (14.86%)		4: 1186 (15.14%)		
	5: 2213 (16.72%)		5: 1546 (19.73%)		

The p value for the difference in variables between the maternal and paternal datasets is included. Data are mean ± standard deviation or number (percentage). The *Age at coronary heart disease (CHD) Event* row represents the mean age at the time of CHD event (hospital admission or death). The *HBSIMD* row represents the number of individuals categorised under each quintile of the Scottish Index of Multiple Deprivation (five meaning least deprived). The *Missing* column refers to the percentage of missing records for each measurement in each dataset. The *Missing* values next to the mean ages of CHD development and death represent the percentage of individuals missing date of birth.

events and died at younger ages than mothers, but had lower deprivation levels on average. The paternal dataset included significantly heavier (13.6g on average) offspring than the maternal dataset, likely due to paternal data being less common during the earlier years of the study, and due to a higher percentage of offspring being male.

Ischaemic heart diseases accounted for the majority of CHD events recorded for both mothers and fathers (Supplementary Table 1). CHD-related hospital admissions accounted for 82.65% and 86.98% of the maternal and paternal CHD events recorded in the dataset, respectively. The characteristics of the dataset used for the supplemental Fine-Gray regression analyses were similar to the main dataset, with no significant differences in the main offspring outcomes studied (Supplementary Table 2).

#### Difference in offspring BW and PW by parental CHD status

Table 2 shows the difference in offspring BW, PW, and PW:BW ratio according to parental CHD status. Offspring from mothers who subsequently developed CHD were, on average, 30.7g lighter (p < 0.001) than offspring from mothers who did not develop CHD. Mothers who developed CHD also had offspring with higher PW:BW ratios, representing higher PW for a given BW (p < 0.001), although PW did not differ. Mothers who developed CHD had shorter pregnancies than those without CHD by around 10 hours (p = 0.021). We found no significant differences in mean offspring BW, PW, PW:BW ratio, or GA between fathers who developed CHD and those who did not.

Fig. 1 shows the difference in offspring BW, PW, and PW:BW ratio by the individual CHD status of each parent. Offspring were born lighter when both parents (p = 0.019) or only the mother (p = 0.011) subsequently developed CHD. In contrast, offspring from parents who both developed CHD had significantly higher PW:BW ratios than offspring from parents who did not develop CHD (p = 0.016) or when only the father did (p = 0.005).

# Parental CHD cumulative incidence by offspring BW and PW quartiles

Fig. 2 shows the cumulative incidence curves for maternal and paternal CHD by offspring BW, PW, and PW:BW ratio quartiles. In mothers, incidence of CHD was significantly higher for those whose offspring was in the lowest BW quartile (Q1), compared to the highest quartile (p < 0.001). Maternal CHD incidence appeared higher for those whose offspring was in the highest PW quartile, but the interquartile differences were not significant. Maternal CHD incidence was significantly higher in those whose offspring PW:BW ratio was in the highest quartile (Q4), compared to the lowest quartile (p < 0.001). The patterns for paternal CHD incidence were similar, but interquartile differences were not significant.

## Cox survival analyses of parental CHD risk

The results from the maternal and paternal survival analyses of CHD risk are shown in Tables 3 and 4, respectively. Parents at the highest risk of developing CHD were those who had offspring born smaller (mothers HR: 0.91, CI: 0.88–0.95, p < 0.001; fathers HR: 0.96, CI: 0.93–1.00, p = 0.048). A decrease of 1 SD in the offspring BW Z-score was associated with a 8.6% and 3.8% increase in the HR for CHD risk in mothers and fathers, respectively. Mothers at higher risk of developing CHD also had higher offspring PW when accounted for BW (HR: 1.14. CI: 1.07–1.22, p < 0.001), and PW:

not Table 2. Summary of variables of interest and their difference between parents who developed coronary heart disease and those who did

	CHD Mothers $(n = 3308)$	Undiagnosed Mothers $(n = 10,558)$	p value	CHD Fathers ( $n = 2967$ )	Undiagnosed Fathers $(n = 5125)$	p value
Offspring						
BW (g)	3352.43 ± 411.65	$3383.17 \pm 412.13$	<0.001	$3384.93 \pm 406.90$	$3392.07 \pm 414.31$	0.450
PW (g)	654.51 ± 114.26	650.81 ± 118.77	0.315	$652.15 \pm 117.11$	$655.21 \pm 121.82$	0.428
PW:BW Ratio	$0.20 \pm 0.03$	$0.19 \pm 0.03$	<0.001	$0.19 \pm 0.03$	$0.19 \pm 0.03$	0.878
GA (weeks)	39.84 ± 1.33	39.90 ± 1.28	0.021	$39.91 \pm 1.28$	39.88 ± 1.27	0.394
Age at Study End (y)	$60.83 \pm 4.56$	59.62 ± 4.69	<0.001	59.49 ± 4.48	58.95 ± 4.51	<0.001
Sex	F: 43.44 %	F: 43.45%	0.988	F: 39.94 %	F: 41.27%	0.240
Parent						
HBSIMD	$2.40 \pm 1.42$	2.64 ± 1.51	<0.001	$2.61 \pm 1.52$	2.76 ± 1.53	<0.001

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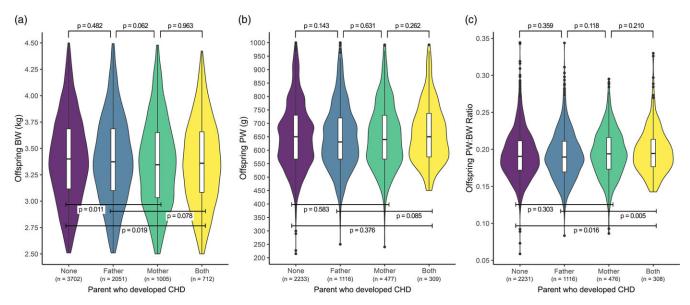


Figure 1. Violin plots of offspring birth weight (A) and placental weight (B) by post-birth development of parental coronary heart disease. Vertical box-and-whiskers plots are included. The p values for the difference in means between each pair of samples (identified by the black lines) was calculated through Welch two sample t-tests.

BW ratio (HR: 1.14, CI: 1.08–1.21, p < 0.001). No significant associations were found between offspring PW or PW:BW ratio and paternal CHD risk.

#### Fine-gray survival analyses of parental CHD risk

In the supplemental Fine-Gray analyses accounting for the competing risk of death (Supplementary Tables 3 and 4), similar associations were found. Mothers at the highest risk of developing CHD were those who had offspring of lower BW (SHR: 0.93, CI: 0.87–0.97, p < 0.001) and higher PW:BW ratio (SHR: 1.26, CI: 1.07–1.49, p = 0.007). No significant associations were found between offspring BW, PW, or PW:BW ratio and the paternal risk of developing CHD accounting for the competing risk of death.

#### **Discussion**

In a novel approach using the Walker cohort, this study investigated the association between parental CHD and offspring birth outcomes. This is the first study including PW measurements and a paternal sample, allowing exploration of associations between offspring PW (and its ratio to BW) and paternal CHD development.

Walker babies from mothers who later developed CHD were born nearly 31g lighter, and this association was independent of whether the father also developed CHD or not. This is in agreement with previous reports of maternal CVD mortality being consistently associated with lower offspring BW.<sup>3,4,7,9,10,14,37</sup> Although we found no difference in offspring BW by whether the fathers subsequently developed CHD or not, using Cox regression we found negative associations between offspring BW and both maternal and paternal CHD risk. This is also supported by the cumulative incidence curves, where the trajectories for parents with offspring in the lowest quartile of BW show increased CHD incidence compared to parents in the highest quartile. Although we identified associations between paternal CHD risk and offspring BW, they are notably of a smaller magnitude than those seen for the mothers, and closer to the 0.05 significance level.

The paternal associations suggest that offspring BW is partially genetically determined through the inheritance of CVD-susceptibility variants. The maternal results, however, are consistent with the strong influence of the intrauterine environment and the reflection of maternal health (and subsequent disease risk) over offspring fetal growth, 38,39 likely obscuring the effect of the maternal genotype. 40,41 One might argue that the paternal associations result simply from the shared parental environment and familial deprivation, but the associations between offspring BW and CVD spread across the extended family, 13,14 strongly suggesting the transmission of CVD-risk alleles through generations. The determination of BW and adult disease has been characterised as a complex mechanism resulting from the interplay between the environmental influences over maternal health (and therefore over the intrauterine environment), the independent expression of the maternal and fetal genomes, and possibly also epigenetic modifications. 14,16,27,37,42

The model including offspring BW and PW showed strong and directionally opposed effect estimates for BW (HR: 0.837) and PW (HR: 1.142), identifying a positive association between the latter and maternal CHD development after accounting for BW. We also found associations between maternal CHD risk and increased offspring PW:BW ratio (HR: 1.144), as a rough reflection of placental inefficiency. This is consistent with the study by Yeung et al.<sup>21</sup> who identified associations between higher offspring PW: BW ratios and increased maternal mortality from several causes such as CVD. Earlier, Davey-Smith et al.5 failed to find such associations, but their sample size was considerably smaller. These results support a link between placental efficiency, fetal growth, and maternal cardiovascular health. Maternal vascular disorders have been linked to poor placental perfusion, leading to insufficiency, impaired fetal growth, and an enlargement of the placenta.<sup>12,43</sup> Increased PW and PW:BW ratios have been associated with adult CVD risk factors and mortality, 17-20 supporting the role of the placenta as a determinant of fetal outcomes and later disease development. We did not find significant associations between paternal CHD risk and offspring PW or PW:BW ratio, but there are no other studies in the literature

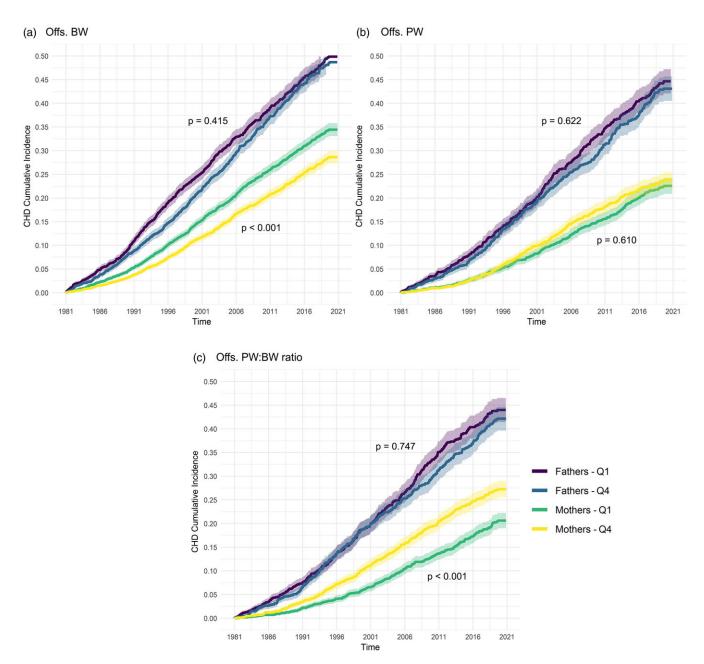


Figure 2. Cumulative incidence curves for parental coronary heart disease by time and offspring (a) birth weight (BW), (b) placental weight (PW), and (c) PW:BW ratio quartiles. Only quartiles 1 and 4 are plotted for clarity. Maternal curves are depicted in green (Q1, lowest quartile) and yellow (Q4, highest quartile). Paternal curves are depicted in purple (Q1) and blue (Q4). The p values for the interquartile difference in trajectories were calculated using Gray's test of equality.

for comparison. In contrast to our previous investigation on parental T2D, <sup>29</sup> we failed to identify paternal associations with offspring PW which could be explained by the fetal inheritance of a disease susceptibility genotype, leading to adult cardiometabolic disorders. Nonetheless, offspring from fathers who later developed CHD had slightly lighter placentas. Alternatively, fetal CVD-susceptibility variants might not be expressed in the placenta, <sup>44</sup> they might be subject to parent-of-origin differential expression, <sup>45</sup> or their action might be outweighed by direct maternal effects.

Our capacity to find a significant negative association between paternal CHD and offspring PW might have been restricted by our late study start point or by the reduced sample with PW available, limiting the statistical power. Through power calculations, we estimated that we had around 50% probability of finding real associations between PW adjusted for BW or PW:BW ratio and paternal CHD of a magnitude similar to those seen in the maternal analyses with the same confidence level (0.001), but around 90% probability of finding them under a 0.05 confidence level. However, these calculations<sup>36</sup> did not incorporate the adjustments performed to comply with the proportional hazards assumption. Recent studies have shown that the association between low BW and the adult development of cardiometabolic disease is governed by a fetal-only effect, being confounded by the maternal effect on the intrauterine environment. <sup>26,46,47</sup> Following this logic and due to the additional evidence hinting at an effect similar to BW, it is likely that our restricted power among other study limitations described below prevented us from finding associations between paternal CHD and offspring PW.

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Table 3. Cox survival analysis of maternal coronary heart disease (CHD) risk

H. R.   C. I. (α = 0.05)   S. E.   P value     Analysis including orly Offs. BW (n = 12,863)     Offs. BW   0.914 (0.881-0.948) 0.019    <0.001     Offs. GA   0.988 (0.961-1.016) 0.014 0.386      Age '81 1.068 (1.063-1.073) 0.002    <0.001     HBSIMD   0.759 (0.688-0.837) 0.050    <0.001     HBSIMD* log(t) 1.048 (1.013-1.083) 0.017 0.006    <0.001     Analysis including orly Offs. PW (n = 5821)     Offs. PW   1.048 (0.990-1.110) 0.029 0.002   <0.007     Age '81 1.113 (1.073-1.154) 0.018    <0.001     HBSIMD   0.736 (0.666-0.894) 0.099 0.002   <0.007     Age '81 1 log(t) 0.94 (1.025-1.169) 0.033 0.007   <0.001     Age '81 1 log(t) 0.984 (0.972-0.996) 0.006 0.012   <0.001     HBSIMD * log(t) 1.031 (0.971-1.095) 0.031 0.312   <0.001     Analysis including Offs. BW and PW (n = 5815)   <0.001     Offs. BW 0.837 (0.780-0.899) 0.036    <0.001     Offs. GA 0.753 (0.620-0.914) 0.099 0.004   <0.001     Offs. GA 1 log(t) 1.096 (1.027-1.170) 0.033 0.006   <0.001     Age '81 * log(t) 0.985 (0.972-0.997) 0.00	Cox survival regression of maternal CHD risk (CHD hospitalisation $+\mbox{death})$						
Offs. BW   0.914   (0.881-0.948)   0.019   <0.001     Offs. GA   0.988   (0.961-1.016)   0.014   0.386     Age '81   1.068   (1.063-1.073)   0.002   <0.001		H. R.	C. I. $(\alpha = 0.05)$	S. E.	P value		
Offs. GA   0.988   (0.961-1.016)   0.014   0.386     Age '81   1.068   (1.063-1.073)   0.002   <0.001	Analysis including only Offs. BW (n = 12,863)						
Age '81   1.068   (1.063-1.073)   0.002   <0.001     HBSIMD   0.759   (0.688-0.837)   0.050   <0.001	Offs. BW	0.914	(0.881-0.948)	0.019	<0.001		
HBSIMD   0.759   (0.688-0.837)   0.050   <0.001     HBSIMD * log(t)   1.048   (1.013-1.083)   0.017   0.006     Analysis including only Offs. PW (n = 5821)               0.017   0.006     Analysis including Offs. PW   1.048   (0.990-1.110)   0.029   0.107   0.002     Age '81   1.113   (1.073-1.154)   0.018   <0.001     HBSIMD   0.781   (0.652-0.936)   0.092   0.007     Offs. GA * log(t)   1.094   (1.025-1.169)   0.033   0.007     Age '81 * log(t)   0.984   (0.972-0.996)   0.006   0.012     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.312     Offs. BW   0.837   (0.780-0.899)   0.036   <0.001     Offs. PW   1.142   (1.068-1.220)   0.034   <0.001     Offs. GA   0.753   (0.620-0.914)   0.018	Offs. GA	0.988	(0.961-1.016)	0.014	0.386		
HBSIMD * log(t)   1.048   (1.013-1.083)   0.017   0.006     Analysis including only Offs. PW (n = 5821)   Offs. PW (n = 5821)     Offs. GA   0.736   (0.606-0.894)   0.099   0.002     Age '81   1.113   (1.073-1.154)   0.018   <0.001     HBSIMD   0.781   (0.652-0.936)   0.092   0.007     Offs. GA * log(t)   1.094   (1.025-1.169)   0.033   0.007     Age '81 * log(t)   0.984   (0.972-0.996)   0.006   0.012     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.312     Analysis including Offs. BW and PW (n = 5815)     Offs. BW   0.837   (0.780-0.899)   0.036   <0.001     Offs. PW   1.142   (1.068-1.220)   0.034   <0.001     Offs. GA   0.753   (0.620-0.914)   0.099   0.004     Age '81   1.113   (1.074-1.154)   0.018   <0.001     HBSIMD   0.789   (0.659-0.945)   0.092   0.010     Offs. GA * log(t)	Age '81	1.068	(1.063-1.073)	0.002	<0.001		
Analysis including only Offs. PW (n = 5821)     Offs. PW   1.048   (0.990-1.110)   0.029   0.107     Offs. GA   0.736   (0.606-0.894)   0.099   0.002     Age '81   1.113   (1.073-1.154)   0.018   <0.001	HBSIMD	0.759	(0.688-0.837)	0.050	<0.001		
Offs. PW   1.048   (0.990-1.110)   0.029   0.107     Offs. GA   0.736   (0.606-0.894)   0.099   0.002     Age '81   1.113   (1.073-1.154)   0.018   <0.001	HBSIMD * log(t)	1.048	(1.013-1.083)	0.017	0.006		
Offs. GA   0.736   (0.606-0.894)   0.099   0.002     Age '81   1.113   (1.073-1.154)   0.018   <0.001	Analysis including only Offs. PW (n = 5821)						
Age '81   1.113   (1.073-1.154)   0.018   <0.001     HBSIMD   0.781   (0.652-0.936)   0.092   0.007     Offs. GA * log(t)   1.094   (1.025-1.169)   0.033   0.007     Age '81 * log(t)   0.984   (0.972-0.996)   0.006   0.012     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.312     Analysis including Offs. BW and PW (n = 5815)     Offs. BW   0.837   (0.780-0.899)   0.036   <0.001	Offs. PW	1.048	(0.990-1.110)	0.029	0.107		
HBSIMD   0.781   (0.652-0.936)   0.092   0.007     Offs. GA * log(t)   1.094   (1.025-1.169)   0.033   0.007     Age '81 * log(t)   0.984   (0.972-0.996)   0.006   0.012     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.312     Analysis including Offs. BW and PW (n = 5815)     Offs. BW   0.837   (0.780-0.899)   0.036   <0.001	Offs. GA	0.736	(0.606-0.894)	0.099	0.002		
Offs. GA * log(t)   1.094   (1.025-1.169)   0.033   0.007     Age '81 * log(t)   0.984   (0.972-0.996)   0.006   0.012     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.312     Analysis including Offs. BW and PW (n = 5815)   0.001   0.001   0.001   0.001     Offs. BW   0.837   (0.780-0.899)   0.036   <0.001	Age '81	1.113	(1.073-1.154)	0.018	<0.001		
Age '81 * log(t)   0.984   (0.972-0.996)   0.006   0.012     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.312     Analysis including Offs. BW and PW (n = 5815)     Offs. BW   0.837   (0.780-0.899)   0.036   <0.001	HBSIMD	0.781	(0.652-0.936)	0.092	0.007		
HBSIMD * log(t)   1.031   (0.971–1.095)   0.031   0.312     Analysis including Offs. BW and PW (n = 5815)     Offs. BW   0.837   (0.780–0.899)   0.036   <0.001     Offs. PW   1.142   (1.068–1.220)   0.034   <0.001	Offs. GA * log(t)	1.094	(1.025-1.169)	0.033	0.007		
Analysis including Offs. BW and PW (n = 5815)   Offs. BW 0.837 (0.780-0.899) 0.036 <0.001	Age '81 * log(t)	0.984	(0.972-0.996)	0.006	0.012		
Offs. BW   0.837   (0.780-0.899)   0.036   <0.001     Offs. PW   1.142   (1.068-1.220)   0.034   <0.001	HBSIMD * log(t)	1.031	(0.971-1.095)	0.031	0.312		
Offs. PW   1.142   (1.068-1.220)   0.034   <0.001     Offs. GA   0.753   (0.620-0.914)   0.099   0.004     Age '81   1.113   (1.074-1.154)   0.018   <0.001	Analysis including Offs. BW and PW ( $n=5815$ )						
Offs. GA   0.753   (0.620-0.914)   0.099   0.004     Age '81   1.113   (1.074-1.154)   0.018   <0.001	Offs. BW	0.837	(0.780-0.899)	0.036	<0.001		
Age '81   1.113   (1.074-1.154)   0.018   <0.001     HBSIMD   0.789   (0.659-0.945)   0.092   0.010     Offs. GA * log(t)   1.096   (1.027-1.170)   0.033   0.006     Age '81 * log(t)   0.985   (0.972-0.997)   0.006   0.014     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.318     Analysis including Offs. PW to BW Ratio (n = 5815)   Compared to the property of	Offs. PW	1.142	(1.068-1.220)	0.034	<0.001		
HBSIMD   0.789   (0.659-0.945)   0.092   0.010     Offs. GA * log(t)   1.096   (1.027-1.170)   0.033   0.006     Age '81 * log(t)   0.985   (0.972-0.997)   0.006   0.014     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.318     Analysis including Offs. PW to BW Ratio (n = 5815)     Offs. PW:BW Ratio   1.144   (1.081-1.210)   0.029   <0.001	Offs. GA	0.753	(0.620-0.914)	0.099	0.004		
Offs. GA * log(t)   1.096   (1.027-1.170)   0.033   0.006     Age '81 * log(t)   0.985   (0.972-0.997)   0.006   0.014     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.318     Analysis including Offs. PW to BW Ratio (n = 5815)   0.029   <0.001	Age '81	1.113	(1.074–1.154)	0.018	<0.001		
Age '81 * log(t)   0.985   (0.972-0.997)   0.006   0.014     HBSIMD * log(t)   1.031   (0.971-1.095)   0.031   0.318     Analysis including Offs. PW to BW Ratio (n = 5815)     Offs. PW:BW Ratio   1.144   (1.081-1.210)   0.029   <0.001	HBSIMD	0.789	(0.659-0.945)	0.092	0.010		
HBSIMD * log(t)   1.031   (0.971–1.095)   0.031   0.318     Analysis including Offs. PW to BW Ratio (n = 5815)     Offs. PW:BW Ratio   1.144   (1.081–1.210)   0.029   <0.001	Offs. GA * log(t)	1.096	(1.027-1.170)	0.033	0.006		
Analysis including Offs. PW to BW Ratio (n = 5815)   Offs. PW:BW Ratio 1.144 (1.081-1.210) 0.029 <0.001	Age '81 * log(t)	0.985	(0.972-0.997)	0.006	0.014		
Offs. PW:BW Ratio   1.144   (1.081-1.210)   0.029   <0.001     Offs. GA   0.742   (0.611-0.901)   0.099   0.003     Age '81   1.113   (1.074-1.154)   0.018   <0.001	HBSIMD * log(t)	1.031	(0.971–1.095)	0.031	0.318		
Offs. GA   0.742   (0.611-0.901)   0.099   0.003     Age '81   1.113   (1.074-1.154)   0.018   <0.001	Analysis including Offs. PW to BW Ratio (n = 5815)						
Age '81 1.113 (1.074-1.154) 0.018 <0.001   HBSIMD 0.783 (0.654-0.938) 0.092 0.008   Offs. GA * log(t) 1.096 (1.027-1.171) 0.033 0.006   Age '81 * log(t) 0.984 (0.972-0.997) 0.006 0.013	Offs. PW:BW Ratio	1.144	(1.081-1.210)	0.029	<0.001		
HBSIMD 0.783 (0.654-0.938) 0.092 0.008   Offs. GA * log(t) 1.096 (1.027-1.171) 0.033 0.006   Age '81 * log(t) 0.984 (0.972-0.997) 0.006 0.013	Offs. GA	0.742	(0.611-0.901)	0.099	0.003		
Offs. GA * log(t) 1.096 (1.027-1.171) 0.033 0.006   Age '81 * log(t) 0.984 (0.972-0.997) 0.006 0.013	Age '81	1.113	(1.074–1.154)	0.018	<0.001		
Age '81 * log(t) 0.984 (0.972-0.997) 0.006 0.013	HBSIMD	0.783	(0.654-0.938)	0.092	0.008		
	Offs. GA * log(t)	1.096	(1.027–1.171)	0.033	0.006		
HBSIMD * log(t) 1.032 (0.972-1.096) 0.031 0.305	Age '81 * log(t)	0.984	(0.972-0.997)	0.006	0.013		
	HBSIMD * log(t)	1.032	(0.972–1.096)	0.031	0.305		

CHD was defined as hospitalisation or death due to CHD. The *H.R.* column represents the hazard ratio for the covariate. The *C.I.* column represents the 95% Confidence Interval for the coefficient. The *S.E.* column represents the regression standard error. *Age '81* refers to the individual 'age in 1981' variable. *HBSIMD* refers to the Scottish Index of Multiple Deprivation (higher meaning less deprived).

For this investigation we have used the Walker cohort, <sup>28</sup> which includes accurate records for three quarters of the total births in the region during its time, and enabling linkage to health records from the parents, which by now have experienced remarkable disease morbidity. However, there are inevitable limitations of such an epidemiological cohort that need to be considered when interpreting our results. In contrast with other reports using a wider cardiovascular mortality outcome definition including a more variable array of conditions, which might be diversely

Table 4. Cox survival analysis of paternal coronary heart disease (CHD) risk

Cox Survival Regression of Paternal CHD Risk (CHD hospitalisation $+$ death)						
	H. R.	C. I. $(\alpha = 0.05)$	S. E.	P value		
Analysis including only Offs. BW (n = 7619)						
Offs. BW	0.962	(0.926-1.000)	0.020	0.048		
Offs. GA	1.022	(0.992-1.053)	0.015	0.146		
Age '81	1.049	(1.044-1.055)	0.003	<0.001		
HBSIMD	0.802	(0.742-0.866)	0.039	<0.001		
HBSIMD * log(t)	1.044	(1.016-1.073)	0.014	0.002		
Analysis including only Offs. PW (n = 3966)						
Offs. PW	0.972	(0.922-1.025)	0.027	0.295		
Offs. GA	1.032	(0.989-1.076)	0.022	0.150		
Age '81	1.072	(1.049-1.096)	0.011	<0.001		
HBSIMD	0.797	(0.709-0.895)	0.059	<0.001		
Age '81 * log(t)	0.992	(0.984-1.001)	0.004	0.068		
HBSIMD * log(t)	1.042	(1.001-1.086)	0.021	0.047		
Analysis including Offs. BW and PW ( $n=3962$ )						
Offs. BW	0.960	(0.900-1.024)	0.033	0.220		
Offs. PW	0.991	(0.932-1.054)	0.031	0.773		
Offs. GA	1.038	(0.994–1.084)	0.022	0.092		
Age '81	1.073	(1.049-1.097)	0.011	<0.001		
HBSIMD	0.798	(0.710-0.896)	0.059	<0.001		
Age '81 * log(t)	0.992	(0.984-1.001)	0.004	0.068		
HBSIMD * log(t)	1.042	(1.001-1.086)	0.021	0.046		
Analysis including Offs. PW to BW Ratio (n = 3962)						
Offs. PW:BW Ratio	1.003	(0.950-1.057)	0.027	0.926		
Offs. GA	1.030	(0.987-1.075)	0.022	0.170		
Age '81	1.072	(1.049-1.096)	0.011	<0.001		
HBSIMD	0.795	(0.708-0.894)	0.059	<0.001		
Age '81 * log(t)	0.992	(0.984-1.000)	0.004	0.063		
HBSIMD * log(t)	1.043	(1.001-1.086)	0.021	0.045		

CHD was defined as hospitalisation or death due to CHD. The *H.R.* column represents the hazard ratio for the covariate. The *C.I.* column represents the 95% confidence interval for the coefficient. The *S.E.* column represents the regression standard error. *Age '81* refers to the individual 'age in 1981' variable. *HBSIMD* refers to the Scottish Index of Multiple Deprivation (higher meaning less deprived).

associated with offspring growth, we have focused on CHD hospitalisations and mortality, for which over 90% were categorised as ischaemic heart diseases. Focusing on CHD only might have limited our sample size. In addition, the hospital admissions data only started being collected after 1980, thirty years after the delivery of the oldest children in the cohort. This carries the possibility of some CHD cases being introduced into the model with a delay or being missed completely due to death or no subsequent admissions during the observable period. This is especially problematic for the paternal analyses due to their earlier onset of CHD. Additionally, the possibility of death before developing CHD should also be considered. The Fine-Gray regression was performed to investigate the associations for CHD

risk accounting for the competing risk of death, but the death records lacked cause of death until 1989, which pushed the study start point further back, increasing the CHD cases missed and therefore being relegated to supplementary material. Any explanatory variables violating the proportional hazards assumption for the Cox and Fine-Gray models were adjusted by adding an interaction term of said covariates with the logarithm of follow-up time. This was considered necessary to prevent proportional hazards violations, albeit increasing the risk of overfitting the model and complicating the interpretation of the results, since the estimated effect of these variables should also consider the time-dependent effect as quantified by the interaction term. The survival regression analyses could not be adjusted for parental weight, height, or smoking and alcohol consumption since these were mostly unavailable for Walker, variables which might act as confounders for offspring outcomes and parental CHD risk. The analyses were adjusted by deprivation index (HBSIMD) from 2019 as a proxy for socio-economic class, which was also missing from Walker. This data resulted highly correlated with a social class categorisation<sup>48</sup> of the parental occupation data available from Walker (analyses not shown). We did not adjust for multiple testing as our separate analyses for BW, PW, and BW:PW as predictors of parental CHD each offered different perspectives on the analyses and were conducted in different subsets of the dataset. Finally, since PW measurements were only collected for the latter half of the cohort, our ability to identify significant associations with offspring PW might have been limited by the analyses being restricted to a younger and smaller parental sample, with lower CHD morbidity. The differences between the entire parental datasets and the subset of individuals for which offspring PW was available are described in Supplementary Table 5.

In conclusion, we have provided additional evidence for independent and directionally opposed effects of maternal CHD risk on offspring BW and PW. Maternal CHD risk was also positively associated with offspring PW to BW ratio, highlighting the importance of placental development and efficiency for fetal growth and its relationship with adult disease. Further analyses with a more comprehensive cohort are required to provide additional insights regarding any association of paternal CHD with offspring PW, and future efforts should focus on unravelling how fetal CVD-susceptibility variants might be influencing not just BW, but placental growth and function.

**Supplementary material.** The supplementary material for this article can be found at https://doi.org/10.1017/S2040174423000430.

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#### Competing interests. None.

**Ethical standards.** Ethical approval for this study was approved through the Health Informatics Centre at the University of Dundee and this study conforms to all recognised standards.

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