Supplementary Information

Estimating the associations between characteristics of BMI trajectories and BP using joint multivariate response models

We applied a joint modelling approach to three response variables: repeated BMI measures (log10 transformed to correct for the right skewness) and SBP and DBP levels at 45y. For the 1958 (both sexes) and 1946 (males) cohorts, we used one knot (t_0) to allow for different "childhood" and "adulthood" BMI curves. For adult SBP and DBP, we specified only an individual-specific (random) intercept (β_{3j} and β_{4j}) to represent BP values for individual j. The joint model log BMI and adult SBP and DBP is written as

$$\log BMI_{ij} = \beta_{0j} + \beta_{1j}t_{ij} + \beta_{2j}(t_{ij} - t_0)I_{t_{ij} > t_0} + e_{ij}$$
 (i)
$$SBP_j = \beta_{3j}$$

$$DBP_j = \beta_{4j}$$

where t_{ij} is the exact age at measurement i. The indicator $I_{t_{ij}>t_0}$ represents two age ranges (=1 for $t_{ij}>t_0$ or =0 otherwise).

For $\log BMI_{ij}$ there are individual-specific intercept β_{0j} , childhood slope β_{1j} (for $t_{ij} \leq t_0$), and adult slope $\beta_{1j} + \beta_{2j}$ (for $t_{ij} > t_0$). Model (i) was fitted by assuming that the random coefficients $\beta_j = (\beta_{0j}, \beta_{1j}, \beta_{2j}, \beta_{3j}, \beta_{4j})^T$ are independently and identically distributed (*iid*) across j and follow a multivariate normal distribution, with mean $\beta = (\beta_0, \beta_1, \beta_2, \beta_3, \beta_4)^T$ and individual-level variance-covariance matrix (unstructured)

$$\Sigma = \begin{bmatrix} \sigma_0^2 & \sigma_{01} & \sigma_{02} & \sigma_{03} & \sigma_{04} \\ & \sigma_1^2 & \sigma_{12} & \sigma_{13} & \sigma_{14} \\ & & \sigma_2^2 & \sigma_{23} & \sigma_{24} \\ & & & \sigma_3^2 & \sigma_{34} \\ & & & & \sigma_4^2 \end{bmatrix}$$

The errors e_{ij} are assumed to be iid across i and j with $e_{ij} \sim N(0, \sigma_e^2)$, and independent of random effects β_j . The three response variables are dependent on each other. The covariance terms

represent the degree to which the individual intercept and slopes for $\log BMI_{ij}$ (β_{0j} , β_{1j} , β_{2j}) covary within themselves and with intercepts (or values) of SBP (β_{3j}) and DBP (β_{4j}).

The individual BMI characteristics of interest include individual-specific $\log BMI_{ij}$ at age t $(\beta_{0j} + \beta_{1j}t + e_{ij} \text{ for } t \le t_0; \ \beta_{0j} - \beta_{2j}t_0 + (\beta_{1j} + \beta_{2j})t + e_{ij} \text{ for } t > t_0)$ and slopes for $\log BMI_{ij}$ $(\beta_{1j} \text{ for } t \le t_0 \text{ and } \beta_{1j} + \beta_{2j} \text{ for } t > t_0)$.

For simplicity, we write $I_{t_{ij}>t_0}=I_0$ and $t-t_0=\Delta t_0$. The correlation between *SBP* and $\log BMI$ is a function of age t:

$$\frac{\sigma_{03} + \sigma_{13}t + \sigma_{23}\Delta t_0 I_0}{\sqrt{[\sigma_0^2 + 2\sigma_{01}t + \sigma_1^2 t^2 + \sigma_2^2 \Delta t_0^2 I_0 + 2(\sigma_{02} + \sigma_{12}t)\Delta t_0 I_0 + \sigma_e^2] \cdot \sigma_3^2}}$$
(1)

The correlation between SBP and each slope $(\beta_{1j} \text{ or } \beta_{1j} + \beta_{2j})$ is written as

$$\frac{\sigma_{13} + \sigma_{23} I_0}{\sqrt{[\sigma_1^2 + (2\sigma_{12} + \sigma_2^2) \cdot I_0] \cdot \sigma_3^2}} \tag{2}$$

The regression coefficient for SBP on log BMI is

$$\frac{\sigma_{03} + \sigma_{13}t + \sigma_{23}\Delta t_0 I_0}{\sigma_0^2 + 2\sigma_{01}t + \sigma_1^2 t^2 + \sigma_2^2 \Delta t_0^2 I_0 + 2(\sigma_{02} + \sigma_{12}t)\Delta t_0 I_0 + \sigma_e^2}$$
(3)

The regression coefficient for SBP on each slope is

$$\frac{\sigma_{13} + \sigma_{23} \cdot I_0}{\sigma_1^2 + (2\sigma_{12} + \sigma_2^2) \cdot I_0} \tag{4}$$

For females in the 1946 cohort, the joint model which includes an additional knot (t_1) is written as

$$\log BMI_{ij} = \beta_{0j} + \beta_{1j}t_{ij} + \beta_{2j}(t_{ij} - t_0)I_{t_{ij} > t_0} + \beta_{2'j}(t_{ij} - t_1)I_{t_{ij} > t_1} + e_{ij}$$

$$SBP_j = \beta_{3j}$$

$$DBP_j = \beta_{4j}.$$
(ii)

The indicator $I_{t_{ij}>t_1}$ takes the value 1 for $t_{ij}>t_1$ or 0 otherwise. In model (ii), the variance-covariance matrix for random coefficients $\boldsymbol{\beta}_j = (\beta_{0j}, \beta_{1j}, \beta_{2j}, \beta_{2j}, \beta_{3j}, \beta_{4j})^T$ is

$$\Sigma = \begin{bmatrix} \sigma_0^2 & \sigma_{01} & \sigma_{02} & \sigma_{02}, & \sigma_{03} & \sigma_{04} \\ & \sigma_1^2 & \sigma_{12} & \sigma_{12}, & \sigma_{13} & \sigma_{14} \\ & & \sigma_2^2 & \sigma_{22}, & \sigma_{23} & \sigma_{24} \\ & & & \sigma_2^2, & \sigma_{2\prime3} & \sigma_{2\prime4} \\ & & & & \sigma_3^2 & \sigma_{34} \\ & & & & & \sigma_4^2 \end{bmatrix}$$

The individual $\log BMI_{ij}$ is written as $\beta_{0j} + \beta_{1j}t + e_{ij}$ for $t \le t_0$, $\beta_{0j} - \beta_{2j}t_0 + (\beta_{1j} + \beta_{2j})t + e_{ij}$ for $t \le t_1$, or $\beta_{0j} - \beta_{2j}t_0 - \beta_{2'j}t_1 + (\beta_{1j} + \beta_{2j} + \beta_{2'j})t + e_{ij}$ for $t \ge t_1$, and three slopes as β_{1j} , $\beta_{1j} + \beta_{2j}$, and $\beta_{1j} + \beta_{2j} + \beta_{2'j}$ respectively.

Let $I_{t_{ij>t_1}}=I_1$ and $t-t_1=\Delta t_1$. The correlation between SBP and $\log BMI$ at age t is written as

$$\frac{\sigma_{03} + \sigma_{13}t + \sigma_{23}\Delta t_0 I_0 + \sigma_{2'3}\Delta t_1 I_1}{\sqrt{\left[\sigma_0^2 + 2\sigma_{01}t + \sigma_1^2 t^2 + 2(\sigma_{02} + \sigma_{12}t)\Delta t_0 I_0 + \sigma_2^2 \Delta t_0^2 I_0 + 2(\sigma_{02}t + \sigma_{12}tt + \sigma_{22}t\Delta t_0)\Delta t_1 I_1 + \sigma_2^2t\Delta t_1^2 I_1 + \sigma_2^2\right] \cdot \sigma_3^2}$$
(5)

The correlation between SBP and each slope $(\beta_{1j}, \beta_{1j} + \beta_{2j}, \text{ and } \beta_{1j} + \beta_{2j} + \beta_{2j})$ can be written as

$$\frac{\sigma_{13} + \sigma_{23} I_0 + \sigma_{2\prime 3} I_1}{\sqrt{\left[\sigma_1^2 + (2\sigma_{12} + \sigma_2^2)I_0 + (2\sigma_{12\prime} + 2\sigma_{22\prime} + \sigma_2^2)I_1\right] \cdot \sigma_3^2}}$$
(6)

The regression coefficient for SBP on log BMI at age t is written as

$$\frac{\sigma_{03} + \sigma_{13}t + \sigma_{23}\Delta t_0 I_0 + \sigma_{2'3}\Delta t_1 I_1}{\sigma_0^2 + 2\sigma_{01}t + \sigma_1^2 t^2 + \sigma_2^2 \Delta t_0^2 I_0 + 2(\sigma_{02} + \sigma_{12}t)\Delta t_0 I_0 + 2(\sigma_{02}\prime + \sigma_{12}\prime t + \sigma_{22}\prime \Delta t_0)\Delta t_1 I_1 + \sigma_2^2 \prime \Delta t_1^2 I_1 + \sigma_e^2} \quad (7)$$

The regression coefficient for SBP on each slope is

$$\frac{\sigma_{13} + \sigma_{23} I_0 + \sigma_{2'3} I_1}{\sigma_1^2 + (2\sigma_{12} + \sigma_2^2) I_0 + (2\sigma_{12'} + 2\sigma_{22'} + \sigma_{2'}^2) I_1}$$
(8)

Models (i) and (ii) were fitted in MLwiN. Model parameters were first estimated using the iterative generalized least squares algorithm which provides maximum likelihood estimates. The knot(s) was

positioned at different ages and -2 log(likelihood) of each model was compared. For model (i), the location of knot t_0 was chosen at 20y for males of both cohorts and at 16y for females of the 1958 cohort based on the likelihood profile (Figures S1a&S2). For model (ii), the knots t_0 and t_1 were chosen at 16 and 34y respectively (Figure S1b). Thus BMI trajectories for females of the 1946 cohort were characterised by three linear curves, with slopes β_{1j} (\leq 16y), $\beta_{1j} + \beta_{2j}$ (16y < t < 34y) and $\beta_{1j} + \beta_{2j} + \beta_{2'j}$ ($t \geq 34y$). The residual plots for the random coefficients β_j in models (i) and (ii) suggest that it is reasonable to assume that they follow a normal distribution (Figures S2 and S3).

The correlations coefficients (or regression coefficients) estimated from Formulae 1-8 are non-linear functions of the model parameters. Their theoretical distributions are complex and the explicit formulae of variances cannot be expressed. We applied the nonparametric bootstrap re-sampling procedure. Each sample was drawn (with replacement) from estimated residuals at each level to create a dataset that has the same multilevel structure. These sampled residuals were then added to the fixed part of the model to obtain a new set of responses. Estimates of BMI levels and childhood and adult slopes and their correlations with BP were obtained for each of the bootstrap sample. We re-sampled 499 bootstrap datasets to obtain the means and 95% CI for the correlations coefficients. Similarly, the correlations between growth characteristics of BMI and DBP can be obtained by replacing the parameters for SBP in Models (i)&(ii) with those for DBP.

Table S1: Regression coefficients; for BP on BMI slopes estimated using joint multivariate models

Males	Systolic BP at 45y			
	1946 cohort		1958 cohort	
	Coefficient	95% CI	Coefficient	95% CI
Child slope (7-20y) †*	0.282	(-0.175,0.793)	1.039	(0.844,1.251)
Adult slope (>20y) †*	1.478	(0.717, 2.269)	2.568	(2.148, 3.097)
Females				
Child slope (7-16y) †	0.594	(0.092, 1.166)	0.920	(0.545,1.321)
Adult slope (16-34y)†*	0.492	(-0.060,1.033)	2.274	(2.035,2.538)
Adult slope (>34y) †*	0.059	(-0.717,0.806)		
Males	Diastolic BP at 45y			
Child slope (7-20y) †	0.648	(0.245, 1.027)	0.623	(0.534, 0.722)
Adult slope (>20y) †	1.657	(1.046, 2.205)	2.021	(1.756,2.371)
Females				
Child slope (7-16y) †	0.455	(0.055, 0.877)	0.607	(0.364, 0.851)
Adult slope (16-34y)†*	0.537	(0.130, 0.939)	1.629	(1.469,1.791)
Adult slope (>34y) †*	0.069	(-0.484, 0.582)		

[†] slope for $\log BMI$ ‡ change in BP for an increase of slope for $\log BMI$ by 0.001 * p<0.05 for test of the difference in regression coefficients between cohorts

Figure S1: Likelihood profile for the 1946 cohort for knots at different ages

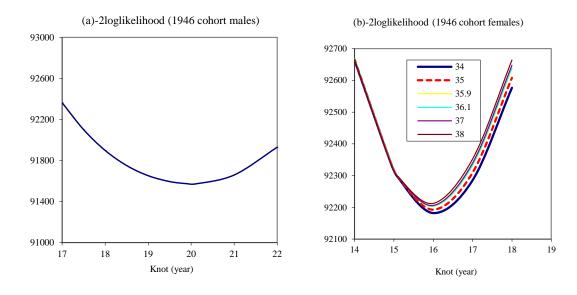
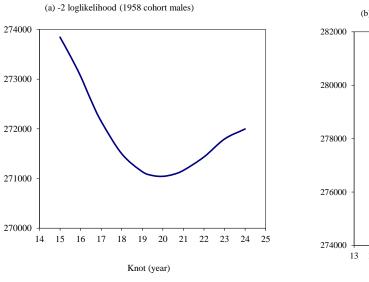


Figure S2: Likelihood profile for the 1958 cohort for knots at different ages



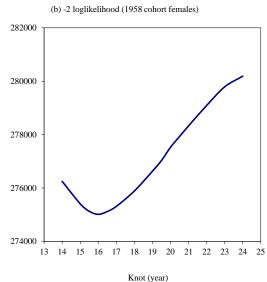


Figure S2: Residual plots for random coefficients in joint multivariate models for the 1946 cohort

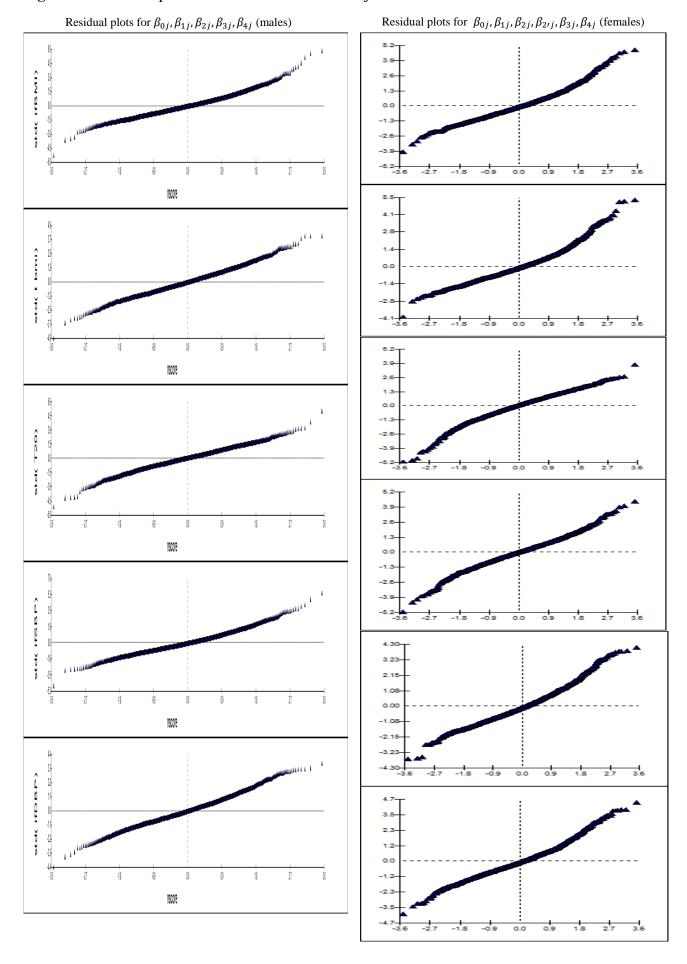


Figure S3: Residual plots for random coefficients in joint multivariate models for the 1958 cohort

Residual plots for $\beta_{0j},\beta_{1j},\beta_{2j},\beta_{3j},\beta_{4j}$ (males) Residual plots for β_{0j} , β_{1j} , β_{2j} , β_{3j} , β_{4j} (females) 5t- 4t- 2t- 4tstd(ifBMI) std(ifBMI) nscore sore std(age) -100 26 307 -100 205 nscore SOUR 21+ 11+ 01-11+ 21+ 41+ std(T20) std(T16) nscore sore std(ifSBP) std(ifSBP) 215 26 nscore SOR 13-55-28std(ifDBP) std(ifDBP) 00 215 190018 Score