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Supplemental Data

Improving Phenotypic Prediction by Combining

Genetic and Epigenetic Associations

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Figure S1. BMI distribution

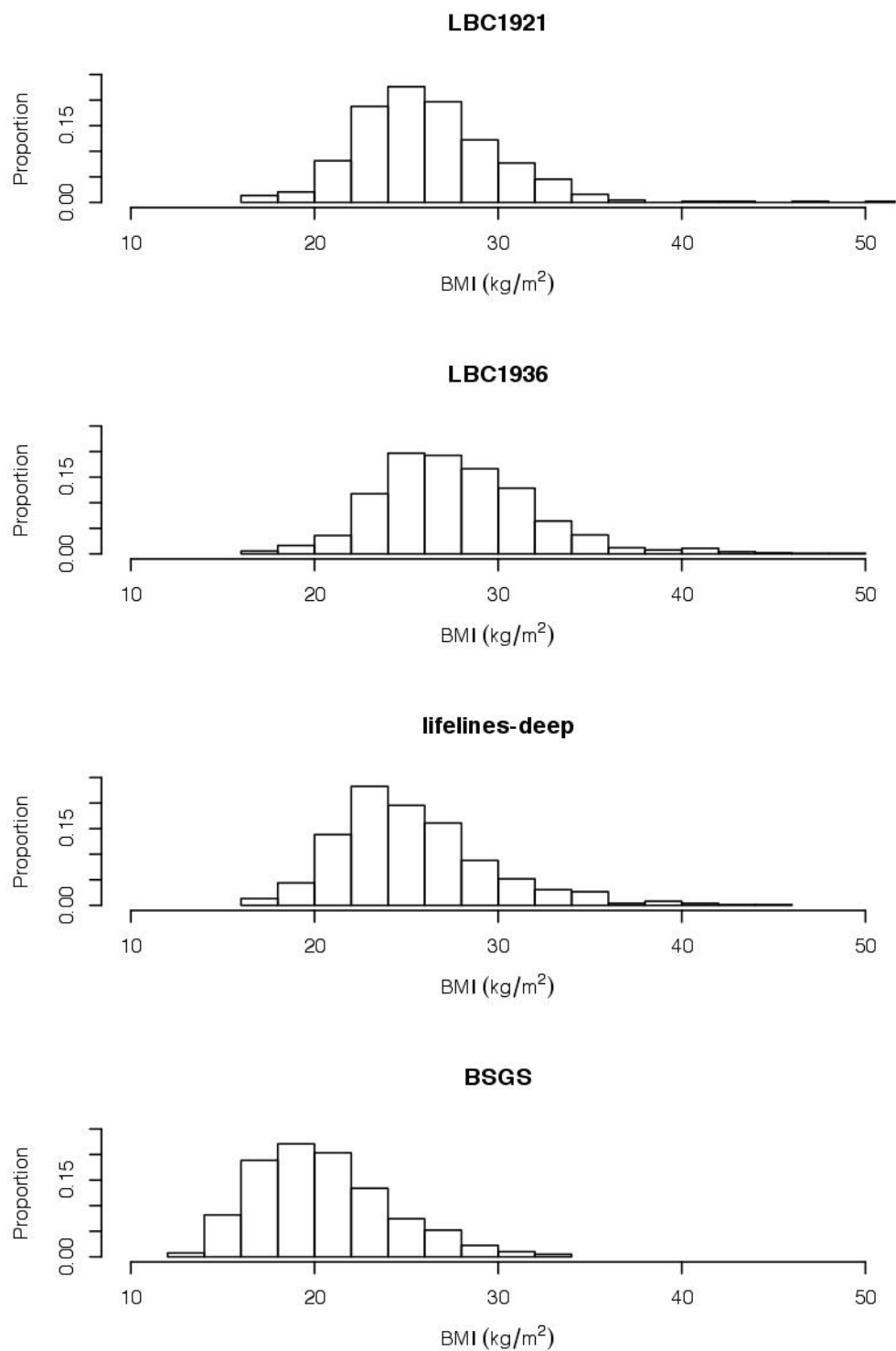


Figure S2. Height distribution

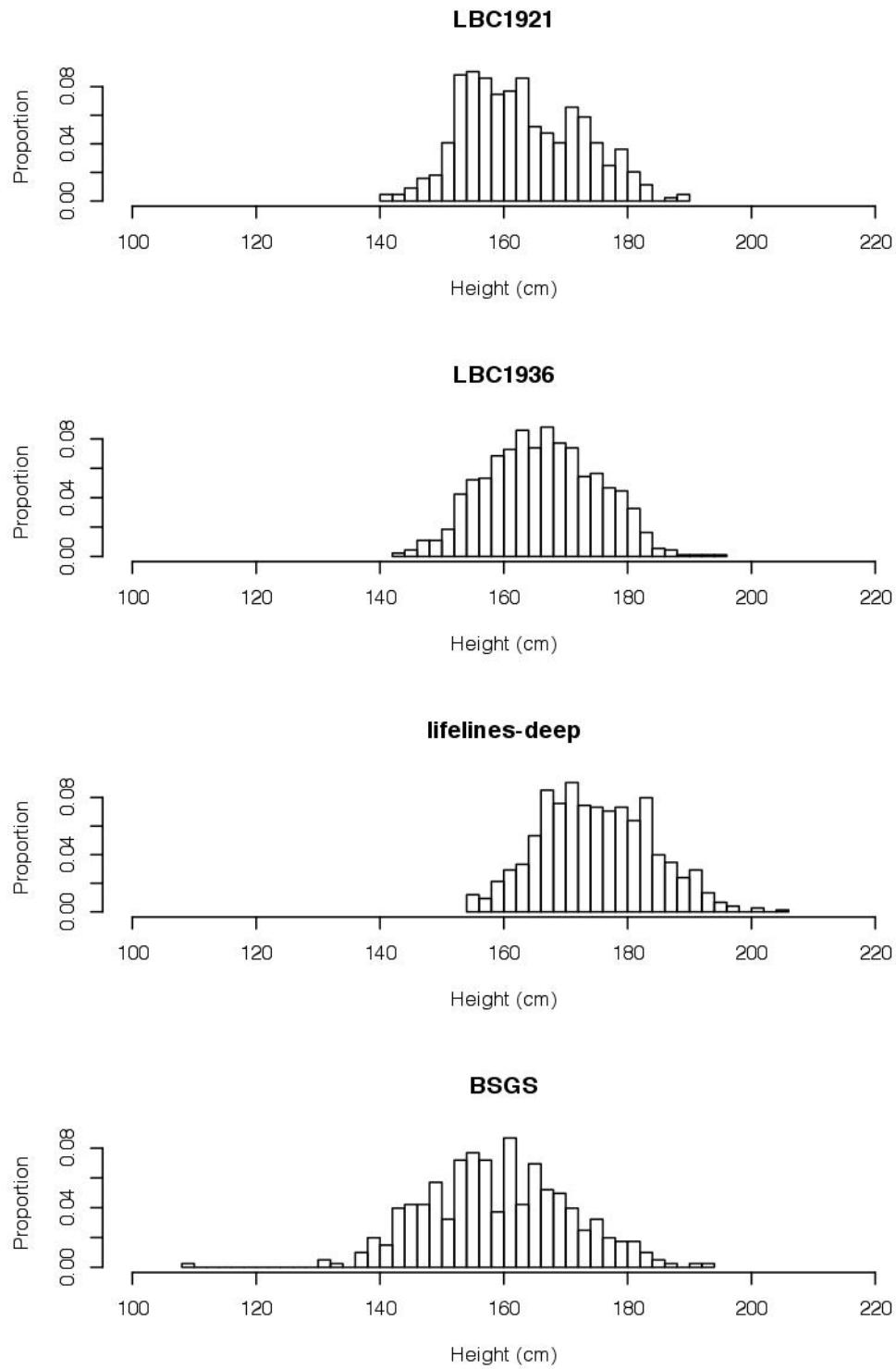


Figure S3. MWAS QQ plots. QQ plots for MWAS for BMI and height in the LBC and Lifelines cohorts.

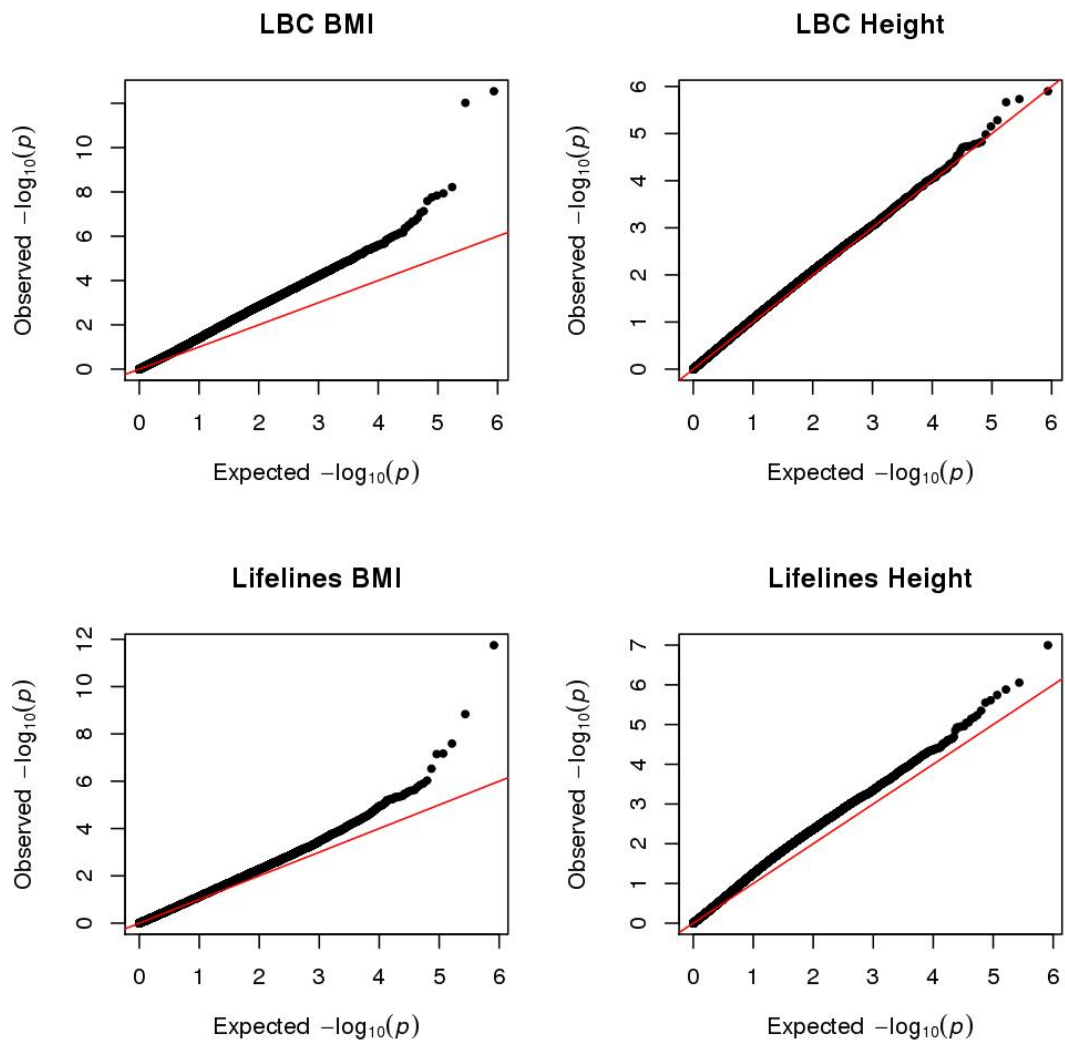


Figure S4. Framingham-based BMI methylation scores. Proportion of variance explained in sex and age-adjusted BMI where the methylation profile scores were based on CpGs identified by the Framingham MWAS. In the LBC, weights for the methylation profile scores were derived from the Lifelines-deep MWAS and *vice versa*. The plot shows the Adjusted- R^2 explained by the methylation profile score, the genetic profile score, an additive model including both scores (Methylation + Genetic), and an interaction model (Methylation * Genetic).

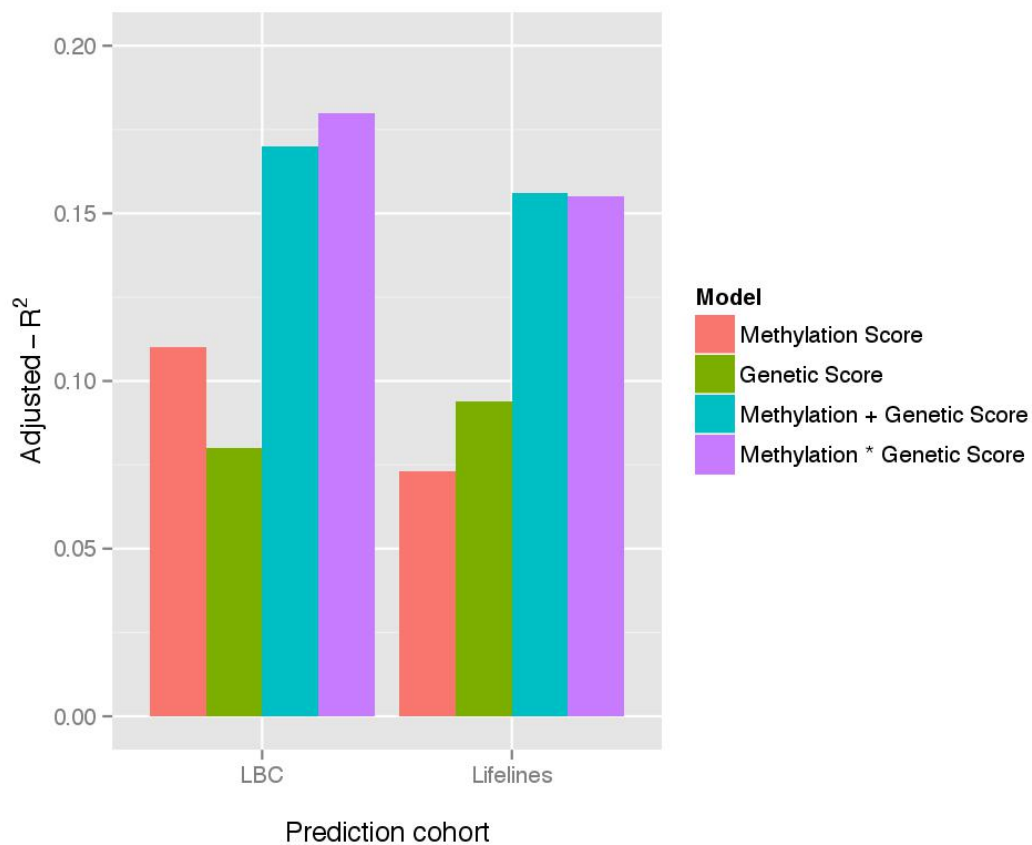


Figure S5. Proportion of variance explained in sex and age-adjusted BMI and height phenotype after correction for cell-count. The plot shows the proportion of variance (Adjusted- R^2) in sex and age-adjusted height explained by the methylation profile score, the genetic profile score, an additive model including both scores (Methylation + Genetic), and an interaction model (Methylation * Genetic).

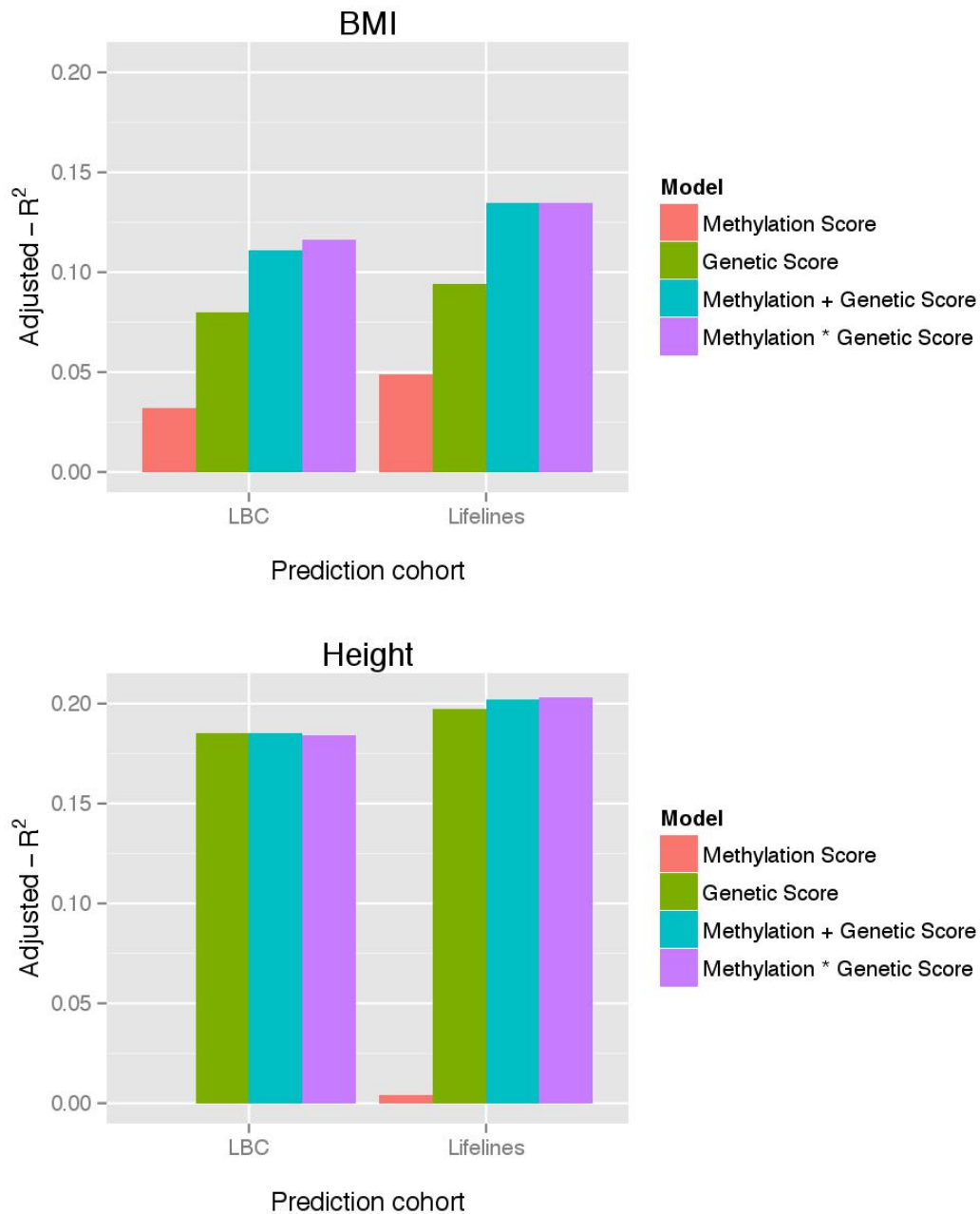


Figure S6. Causality and reverse causation.

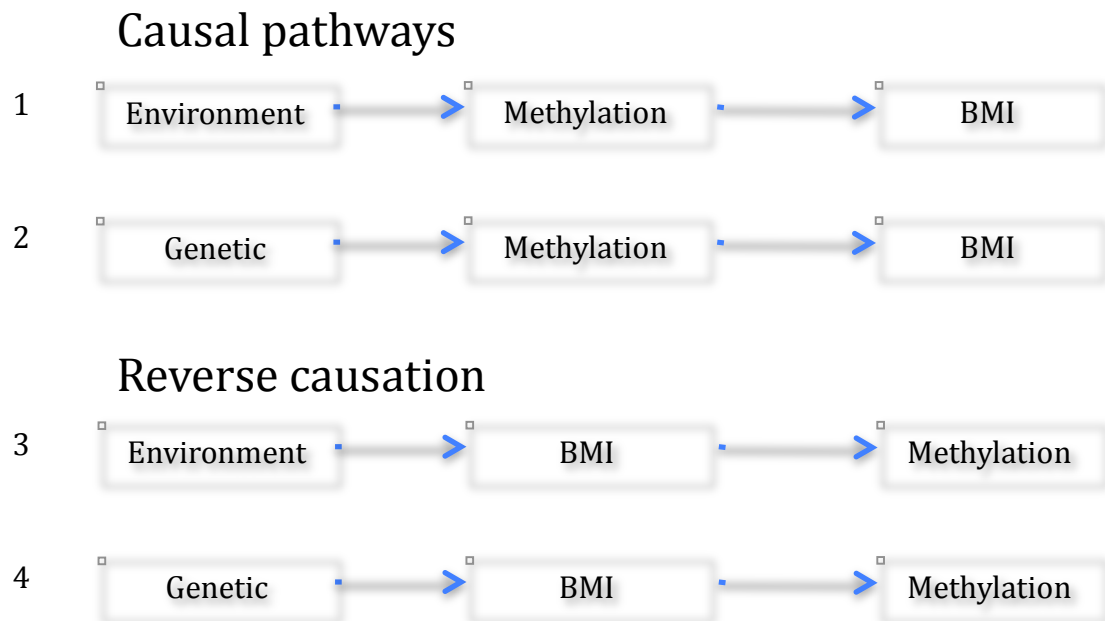


Table S1. CpG probes significantly associated with BMI in LBC and Lifelines-deep. Effect sizes are per standard deviation change in sex and age-adjusted BMI with a unit change in the transformed methylation values.

Probe	chr	BP (NCBI37)	R ²	effect	SE	P
LBC						
cg06500161	21	43656587	0.038	0.196	0.027	2.85E-13
cg11024682	17	17730094	0.036	0.192	0.027	9.56E-13
cg07094298	4	2748026	0.024	-0.157	0.027	6.04E-09
cg04011474	2	28904455	0.023	-0.154	0.027	1.16E-08
cg10192877	21	43641690	0.023	0.153	0.027	1.45E-08
cg17782974	10	104406990	0.022	0.152	0.027	1.76E-08
cg20496314	22	39759864	0.022	0.150	0.027	2.57E-08
cg07202479	1	159174162	0.020	-0.145	0.027	7.35E-08
cg09494176	4	3193868	0.020	-0.144	0.027	9.04E-08
Lifelines						
cg11024682	17	17730094	0.060	0.253	0.035	1.79E-12
cg17501210	6	166970252	0.045	-0.218	0.036	1.46E-09
cg21139312	17	55663225	0.038	0.201	0.036	2.57E-08
cg06500161	21	43656587	0.035	0.195	0.036	6.75E-08
cg26403843	5	158634085	0.036	0.195	0.036	7.14E-08

Table S2. Correlation between sex- and age-adjusted BMI and height with methylation and genetic predictors

Methylation Predictor						
Trait	Probe selection cohort	Effect size estimation cohort	Prediction cohort	Pearson Correlation	Adjusted-R ²	Pearson correlation P-Value
BMI	Lifelines	Lifelines	LBC	0.26	0.069	<1x10 ⁻¹⁵
BMI	LBC	LBC	Lifelines	0.22	0.049	6.685x10 ⁻¹⁰
BMI	Framingham	Lifelines	LBC	0.33	0.11	<1x10 ⁻¹⁵
BMI	Framingham	LBC	Lifelines	0.27	0.073	3.952x10 ⁻¹⁴
Height	Lifelines	Lifelines	LBC	0.06	0.0031	0.02
Height	LBC	LBC	Lifelines	0.09	0.0076	0.0096
Genetic Predictor						
Trait	Probe selection cohort	Effect size estimation cohort	Prediction cohort	Pearson Correlation	Adjusted-R ²	Pearson correlation P-Value
BMI	GIANT 2015	GIANT 2015	LBC	0.28	0.08	<1x10 ⁻¹⁵
BMI	GIANT 2015	GIANT 2015	Lifelines	0.31	0.094	<1x10 ⁻¹⁵
Height	GIANT 2014	GIANT 2014	LBC	0.43	0.19	<1x10 ⁻¹⁵
Height	GIANT 2014	GIANT 2014	Lifelines	0.45	0.20	<1x10 ⁻¹⁵

Table S3. Additive vs interaction model. Proportion of variance explained in sex- and age-adjusted BMI and height by the additive genotype/methylation model versus an interaction model.

Trait	Methylation Predictor		Prediction cohort	Model	R ²	ANOVA p-value for model comparison
	Probe selection cohort	Effect size estimation cohort				
BMI	Lifelines	Lifelines	LBC	Additive	0.14	5.0x10⁻⁶
				Interaction	0.15	
BMI	LBC	LBC	Lifelines	Additive	0.13	0.35
				Interaction	0.13	
BMI	Framingham	Lifelines	LBC	Additive	0.17	1.4x10⁻⁴
				Interaction	0.18	
BMI	Framingham	LBC	Lifelines	Additive	0.16	0.59
				Interaction	0.16	
Height	Lifelines	Lifelines	LBC	Additive	0.19	0.64
				Interaction	0.19	
Height	LBC	LBC	Lifelines	Additive	0.20	0.50
				Interaction	0.20	

Table S4. Relative contribution of methylation and genetic scores to variance in BMI. The table shows the relative contribution of the methylation and genetic profile scores to the variation in the sex and age-adjusted BMI phenotype (the ratio of the R^2 values from the methylation and genetic profile score) in the adult cohorts compared to BSGS adolescents individuals.

Methylation score derivation	Prediction cohort	Ratio of Methylation score R^2 /genetic score R^2
Discovery cohort=LBC; weights=LBC	Lifelines	0.517
Discovery cohort=LBC; weights=LBC	BSGS	0.015
Discovery cohort=Framingham; weights=LBC	Lifelines	0.777
Discovery cohort=Framingham; weights=LBC	BSGS	0.583
Discovery cohort=Framingham; weights=Lifelines	LBC	1.375
Discovery cohort=Framingham; weights=Lifelines	BSGS	1.056
Discovery cohort=Lifelines; weights=Lifelines	LBC	0.863
Discovery cohort=Lifelines; weights=Lifelines	BSGS	0.694

Table S5. Effect of p-value threshold on prediction ability of methylation score

Trait	Prediction Cohort	P-value threshold	Adj R2	P-value
BMI	LBC	0.1	0.001	0.09
	LBC	0.001	-0.0005	0.5
	LBC	1.00E-05	0.037	6×10^{-13}
	LBC	1.00E-07	0.069	$<1 \times 10^{-15}$
	Lifelines	0.1	-0.00005	0.3
	Lifelines	0.001	-0.0006	0.5
	Lifelines	1.00E-05	-0.0007	0.5
	Lifelines	1.00E-07	0.049	7×10^{-10}
Height	LBC	0.1	0.002	0.0407
	LBC	0.001	0.003	0.02161
	LBC	1.00E-05	0.0003	0.225
	LBC	1.00E-07	NA	NA
	Lifelines	0.1	0.0058	0.02
	Lifelines	0.001	0.0076	0.01
	Lifelines	1.00E-05	-0.0001	0.3
	Lifelines	1.00E-07	NA	NA