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1 **Change in clusters of lifestyle behaviours from childhood to adolescence: a**
2 **longitudinal analysis.**

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10 were involved in data collection and management of the study.

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11 **Abstract**

12 **Purpose:** This study aimed to identify changes in clusters of lifestyle behaviours (physical activity,
13 screen time and diet) between the ages of 7 and 14 years, and to examine socio-demographic determinants
14 of changes.

15 **Methods:** Longitudinal analyses were performed on a sample of 9,339 children from the UK Millennium
16 Cohort Study (MCS) who had complete data on behaviours of interest at age 7 (wave 4) and 14 years
17 (wave 6). Joint Correspondence Analysis (JCA) and k-means cluster analysis were used to identify
18 clusters of lifestyle behaviours at both time waves. Multinomial logistic regressions were used to examine
19 the associations between socioeconomic variables and changes in cluster membership. Analyses were
20 conducted separately for boys and girls.

21 **Results:** Clusters of behaviours at age 7 and 14 years were identified as healthy, mixed or unhealthy
22 respectively. Compared to girls, a higher proportion of boys remained in the healthier cluster over time
23 (19.1% vs. 13.1%) or became healthier (26.4% vs. 9.36%). A higher proportion of girls changed to an
24 unhealthier cluster (57.2% vs. 33.9%). Indicators of lower socio-economic status, such as low family
25 income, low parental education, and not living with both parents at age 7 were associated with unhealthier
26 changes in cluster membership.

27 **Conclusion:** Lifestyle behaviours cluster in children and are susceptible to change over a 7-year period,
28 with a high proportion of boys becoming healthier and a higher proportion of girls became unhealthier.
29 Indicators of socio-economic status appear to be important in determining changes in clusters.

30 **Keywords:** lifestyle behaviours, clusters, longitudinal, children, adolescents, inequalities.

31 **What is known:** poor lifestyle behaviours (i.e. unhealthy dietary habits, low physical activity, and
32 sedentary behaviours) tend to cluster in children and adolescents.

33 **What is new:** Lifestyle behaviours cluster in children and are susceptible to changes between childhood
34 and adolescence. Changes occur differently in boys and girls. Indicators of low socio-economic status are
35 associated with unhealthier changes in behavioural clusters.

36 **List of abbreviations:** HC: Healthy Cluster, IQR: interquartile range, JCA: Joint Correspondence
37 Analysis, OECD: Organisation for Economic Co-operation and Development, MC: Mixed Cluster, MCA:
38 Multiple Correspondence Analysis, MCS: Millenium Cohort Study, MVPA: moderate-to-vigorous

39 physical activity, NVQ: National Vocational Qualification, OR: odds ratio, TV: television, UC:
40 Unhealthy Cluster.

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41 **Introduction**

42 Lifestyle behaviours such as diet, physical activity, and sedentary behaviour have a significant impact on
43 the health and wellbeing of children and young people [1, 2]. Previous studies suggest that the transition
44 from childhood to adolescence is a critical period for behaviour formation [3, 4]. Adolescence is often
45 accompanied by increases in screen time, decreases in physical activity, and poorer dietary behaviours,
46 characterised by a low consumption of fruit and vegetables, higher consumption of snacks and sugar-
47 sweetened beverages, and breakfast skipping [3, 5–8]. The acquisition of poor lifestyle behaviours during
48 adolescence can result in higher adiposity and cardiometabolic risk [9–11], lower cardio-respiratory
49 fitness [9, 11], poorer self-rated health [12], poorer mental and emotional health [9, 11, 13], and lower
50 quality of life [10].

51 Much of the evidence on health behaviours of adolescents comes from studies looking at individual health
52 behaviours in isolation. However, lifestyle behaviours are likely to co-occur within groups of young
53 people [9, 14]. For example, it has been shown that high screen time and high energy-dense snacks
54 consumption tend to cluster in young people, as does low fruit consumption and low physical activity [9,
55 14]. The effect of clusters of behaviours on health is likely to be greater than the effect of individual
56 behaviours [15, 16]. A better understanding of how, and in whom lifestyle behaviours cluster, and how
57 clusters of behaviours evolve over critical life periods of childhood is essential for the development of
58 targeted interventions. However, to date, much of the evidence on clustering of health behaviours has
59 come from cross-sectional studies [9, 17] or research that has not examined changes in clusters from
60 childhood to adolescence [9]. Previous research has shown that a high proportion of 5-6 years old
61 children belonged to an unhealthy behavioural cluster (high sedentary and low physical activity, or high
62 TV and high energy-dense snacks consumption) and tended to remain in the same cluster or move to the
63 other unhealthy cluster by age 9 years [18]. Furthermore, in a study of adolescents aged 14-16 years at
64 baseline there was a high stability in clusters of lifestyle behaviours (healthy and unhealthy) over a two-
65 year period [19]. However, it is unknown how clusters of lifestyle behaviours change as children become
66 adolescence. Given the significant social, environmental, and interpersonal changes that go alongside this
67 transition [3, 20], it is imperative to understand in whom and how clusters of behaviours change during
68 this period to help inform public health interventions.

69 An understanding of the determinants of behavioural clusters is required to underpin public health
70 interventions aimed at promoting positive changes in health behaviours [17]. Given that behaviour change

71 interventions that are theoretically based have been shown to be more effective than those without a
72 theoretical underpinning [21, 22], it is useful to utilise behavioural theories to provide a framework for
73 studying determinants of behavioural clusters. There is support for the use of socio-ecological models for
74 understanding key determinants of health behaviours [23]. The socio-ecological model acknowledges that
75 children are embedded within larger social systems and that characteristics of children interact with
76 environments that underlie health behaviours [24].

77 Previous research has identified that indicators of socio-economic position, factors from a most pertinent
78 level of the socio-ecological model, are important influences on children's behavioural clusters, partly
79 because they influence children's attitudes, experiences, behaviours and exposures to health risk factors
80 [25]. For example, cross-sectional studies have shown that higher parental education is associated with
81 behavioural clusters including high physical activity in schoolchildren [26], while girls and children from
82 lower educated parents have a higher prevalence of unhealthier clusters (for example, clusters including
83 high consumption of sugar sweetened beverages, high screen time, and low sleep duration) [26, 27].
84 However, wider socio-economic determinants of changes in clusters of lifestyle behaviours are poorly
85 understood.

86 The results from this study provide support for intervening on multiple behaviours during childhood and
87 adolescence, and to focus on gender differences and parental characteristics related to socioeconomic
88 disparities. More longitudinal analyses in different countries that help to better understand how young
89 boys and girls from different socioeconomic positions change their lifestyle behaviours over the time may
90 guide research to develop more specific and targeted interventions that better consider the needs of
91 different subgroups of young people.

92 **Methods**

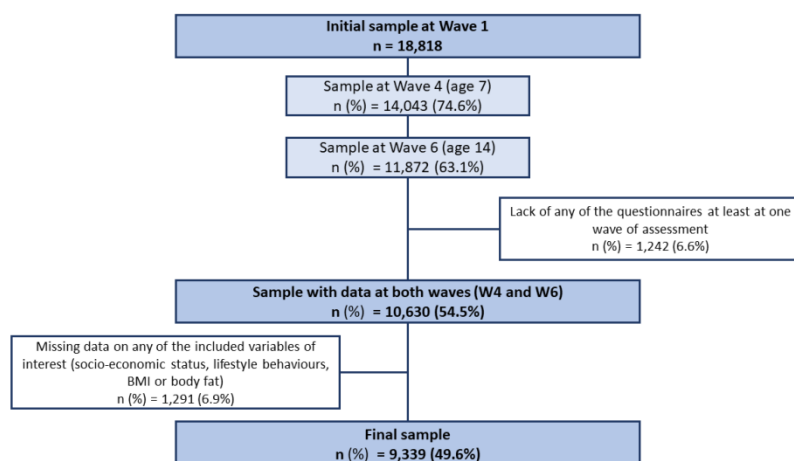
93 *Study design and participants*

94 Data from the UK Millennium Cohort Study (MCS) ([https://cls.ucl.ac.uk/cls-studies/millennium-cohort-](https://cls.ucl.ac.uk/cls-studies/millennium-cohort-study/)
95 [study/](https://cls.ucl.ac.uk/cls-studies/millennium-cohort-study/)) were used for the present analyses. Details of the design and sampling of MCS has been published
96 elsewhere [28]. In brief, the MCS is an observational nationally representative longitudinal birth cohort
97 study, which examines socio-economic, and health related circumstances of children born in the UK
98 between September 2000 and January 2002. At baseline (Wave 1) 18,818 children were first assessed
99 when they were 9 months of age. For the present analyses, data when participants were aged 7 years

100 (MCS4: January-December 2008; n = 14,043) and aged 14 years (MCS6: January 2015-March 2016; n =
 101 11,872) were used (Figure 1).

102 **Figure 1. Sample selection flow-chart**

103



104

105 *Measures*

106 All assessments of behaviours (physical activity, screen-time, and dietary behaviours) were collected via
 107 parent report when participants were aged 7 years, and via self-report when participants were aged 14
 108 years. For the present analyses, variables were dichotomized considering the value which were closer to
 109 the guideline recommendations [29, 30] and the sample distribution.

110 *Physical activity*

111 Physical activity was reported as the number of days per week that the cohort member (child participant)
 112 took part in moderate-to vigorous physical activity (MVPA). For the present analysis, MVPA was
 113 classified as '< 5 days/week' and '≥ 5 days/week'.

114 *Screen time*

115 The number of hours spent watching television (TV) and playing electronic games on a typical weekday
 116 was reported. Time spent TV viewing was dichotomized into '< 3 hours/day' and '≥ 3 hours/day', and
 117 time spent using electronic games was dichotomized into '< 1 hour/day' and '≥ 1 hour/day'.

118 *Dietary behaviours*

119 Participants reported the servings of fruit eaten per day, and the number of days per week breakfast was
120 eaten. Fruit consumption was dichotomized into '< 2 servings/day' and '≥ 2 servings/day', and the
121 frequency of eating breakfast was dichotomized into 'not every day' and 'everyday'.

122 *Socioeconomic variables at age 7 years*

123 When participants were aged 7 years, parents reported their child's sex, ethnicity, and their weekly
124 household income according to the Organisation for Economic Co-operation and Development (OECD),
125 participants whose OECD was below 60% of the median were considered in situation of poverty. Parents
126 also reported their highest National Vocational Qualification (NVQ) level achieved, and who the child
127 lived with at home.

128 *Statistical analyses*

129 Analyses were conducted using Stata software version 14 (Stata, College Station, TX). Baseline sample
130 characteristics were computed by sex and summarized as median and interquartile range (IQR) for
131 continuous variables, and frequencies and percentages for categorical variables. Kolmogorov Smirnov
132 test and histograms were used to assess the variable distribution. Mann Whitney U tests were used for sex
133 differences in continuous non-parametrical variables, and chi-square tests was used for sex differences in
134 proportions.

135 At both time points, Joint Correspondence Analysis (JCA), which is an extension of the Multiple
136 Correspondence Analysis (MCA), was conducted to explore the patterns in the categorical lifestyle
137 variables. The analysis included five variables (physical activity, TV viewing, electronic games playing,
138 breakfast eating and fruit consumption). Analyses were carried out for boys and girls separately, as
139 significant sex differences were shown in the lifestyle behaviours at baseline (Table 1). The dimensions
140 revealed from the JCA that contained most of the data variability were used to inform the cluster analysis.
141 The Calinski-Harabasz rule [31] was used to determine the optimal number of clusters solutions. The
142 final number of clusters was decided based on interpretability and Calinski-Harabasz indexes at both time
143 points and for both sexes. So, the number of clusters less or equal to five, with a higher Calinski-Harabasz
144 index at both time points and for both sexes were selected. K-means cluster analysis was used to create 3
145 clusters of behaviours using the coordinates obtained from the JCA at both waves of assessment
146 independently, and for boys and girls separately. The stability (tracking) of cluster membership between
147 ages 7 years and 14 years was examined using Kappa (K) statistics. The degree of tracking was evaluated

148 using the cut-offs suggested by Landis and Koch [32]: < 0.2, poor; 0.2–0.4, fair; 0.41–0.6, moderate;
149 0.61–0.8, good; > 0.81, very good. The K value for the stability of cluster membership between 7 and 14
150 years was 0.254 for girls and 0.265 for boys, which shows fair agreement between baseline and follow-up
151 cluster membership.

152 As the tracking of clusters was fair, we determined the profile of changes in behavioural clusters.
153 Participants were classified depending on their continuity in the same cluster over time or by their change
154 from one cluster to another. The categories to represent change between age 7 and 14 years were
155 ‘remained in healthy cluster’, ‘remained in mixed cluster’, ‘remained in unhealthy cluster’, ‘changed to an
156 unhealthier cluster (i.e. moved from healthy at age 7 to mixed or unhealthy at age 14; or from mixed at
157 age 7 to unhealthy at age 14)’, and ‘changed to a healthier cluster (i.e. moved from unhealthy at age 7 to
158 mixed or healthy at age 14; or from mixed at age 7 to healthy at age 14)’. Multinomial logistic regressions
159 and the calculation of the corresponding odds ratios (OR) were used to examine the associations between
160 socioeconomic variables at baseline and the changes in cluster membership between ages 7 and 14 years.
161 These models were performed for boys and girls separately and were adjusted for weight status and body
162 fat percentage. Anthropometric covariates were collected by trained researchers during home visits when
163 children were aged 7 years (MCS 4).

164 **Results**

165 From the initial MCS sample of 18,818 children, a final sample analytical sample of 9,339 children
166 (4,660 girls and 4,679 boys; 49.6% of the initial sample) had complete data on all variables of interest and
167 were included in the present analyses (Figure 1).

168

169 *Characteristics of participants at age 7 years*

170 Parents of boys had a high education level (NVQ above 3) (40.27% vs. 38.65%; $p = 0.035$) and had
171 higher weekly family income (£405.27 ± 232.47 vs. £397.99 ± 234.50; $p = 0.038$) compared to girls
172 (Table 1).

173 At age 7, girls participated in MVPA less frequently and skipped breakfast more frequently than boys.

174 Boys consumed less servings of fruit and spent more time watching TV and playing electronic games
175 than girls (Table 1).

176 ***Characteristics of Behavioural Clusters***

177 For boys, the two dimensions (x and y) revealed by the JCA analyses explained 76.1% and 26.1% of the
178 variability in the data at baseline, and the 89.4% and 10.4% variability in the data at follow-up. For girls,
179 the two dimensions revealed by the JCA analyses explained 82.4% and 17.5% of the variability in the
180 data at baseline, and the 86.5% and 13.4% variability in the data at follow-up (SF 1).

181 Three behavioural clusters solutions were produced: a Healthy Cluster (HC), a Mixed Cluster (MC) and
182 an Unhealthy Cluster (UC) (Table 2). At baseline, 41.6% of boys and 52.6% of girls were in the HC,
183 36.5% of boys and 31.2% of girls were in the MC, and 21.9% of boys and 16.2% of girls were in the UC.

184 Among boys and girls at baseline, the HC was characterized by no breakfast skipping, lower TV viewing,
185 higher MVPA, higher fruit consumption, and lower e-games playing than in the MC and UC (Table 2).
186 The UC was characterized by lower MVPA, higher breakfast skipping, lower fruit consumption
187 (especially in boys), higher TV viewing, and higher e-games playing than HC and MC. The MC was
188 characterized by lower MVPA, low breakfast skipping, low TV watching higher, high electronic games
189 playing, and fruit consumption lower than HC, but higher than UC (Table 2).

190 At follow up, there were differences in the behaviours that characterised each cluster, and clusters
191 differed more between boys and girls than at baseline (Table 2). In boys, the three clusters had a similar
192 percentage of high e-games playing (between 28.8% and 37.4%), while 76% of the girls with high e-
193 games playing were in the UC. In girls, the MC was characterized by high fruit consumption (similar to
194 HC), while in boys the MC was characterized by low fruit consumption (Table 2).

195 ***Changes in behavioural clusters between age 7 and 14 years***

196 Almost a fifth of boys (19.1%) who were in the HC at baseline remained in the HC at age 14 compared
197 with 13.1% of girls (Table 3). A similar proportion of boys and girls who were in the MC at age 7
198 remained in the MC at age 14 years (11.2% and 9.1%), and a higher proportion of boys (7%) who were in
199 the UC at age 7 remained in the UC at age 14 years compared to girls (3.4%). A high percentage of boys
200 (26.4%) became healthier compared to girls (9.4%). A higher proportion of girls became unhealthier
201 compared to boys (57.2% vs. 33.9%).

202 *Associations between indicators of socio-economic and change in behavioural cluster profiles.*

203 After controlling for socio-economic indicators and anthropometric factors, compared to the boys who
204 remained in HC, boys from low-income households were more likely to change to a healthier cluster than
205 those from a higher income household (OR (95%CI) = 1.43 (1.10, 1.84)) (Table 4). Boys with parents of
206 higher education were less likely to remain in the MC, remain in the UC and change to an unhealthier
207 cluster, but also less likely to change to a healthier cluster than those of less educated parents (NVQ
208 below 3 or other qualifications). Boys living with a single parent or other guardian were more likely to
209 remain in MC, remain in UC and change to an unhealthier cluster, but also less likely to change to a
210 healthier cluster compared to those living with both parents.

211 Compared to the girls who remained in HC, girls from lower income households were more likely to
212 remain in UC and change to an unhealthier cluster (Table 5). Girls with higher educated parents were less
213 likely to remain in MC, remain in UC, and change to an unhealthier cluster, but also less likely to change
214 to a healthier cluster than those with lower parental education. Girls whose parents had other
215 qualifications were two times more likely to remain in UC (OR (95%CI) = 2.25 (1.40, 3.62)), and two
216 times more likely to change to a healthier cluster (OR (95%CI) = 2.14 (1.30, 3.52)) compared to girls
217 who parents have lower education levels. Girls from White ethnic groups were less likely to remain in
218 MC and less likely to change to a healthier cluster compared to girls from other ethnic groups. Girls living
219 with a single parent or other guardian were more likely to remain in UC and change to an unhealthier
220 cluster compared to those living with both parents.

221 **Discussion**

222 The aims of this study were to identify the changes in behavioural clusters from childhood to adolescence
223 and to examine the associations between indicators of socio-economic status and changes in behavioural
224 clusters over the time. To the best of our knowledge these are the first analyses to have examined changes
225 in clusters of health behaviours over 7 years, to have considered the transition from childhood to
226 adolescence, and associated these changes to socio-economic indicators. Findings show low consistency
227 in behavioural clusters over the 7 years age 7 and 14 years with clear sex differences, as girls tend to
228 change to unhealthier clusters, while boys tend to remain in healthier clusters. Indicators of socio-
229 economic status during childhood appear to be important determinants of changes in behavioural clusters
230 over the 7-year period.

231 The present study found that behavioural clusters are susceptible to changes over a 7-year period. At age
232 7 years boys and girls had similar cluster prevalences, being in the HC was the most prevalent and in the
233 UC was the least prevalent. This changed in boys to a an almost equal prevalence across the three
234 clusters, whereas among girls, the UC became the most prevalent and the HC the least prevalent at age 14
235 years. Previous studies have also shown that clusters of health behaviours have low to moderate tracking
236 rates from ages 5-11 and 9-14 years [18]. Within the clusters in the present study, the number of boys
237 with high e-games playing increased over the time and was similarly distributed among the three clusters,
238 and the number of boys and girls with high TV watching increased, especially in the UC. These results
239 are in line with other studies reporting an increase in screen time from childhood to early adolescence,
240 with differences in media used among boys and girls [33], and such increases in screen time could be
241 driving the changes seen in certain clusters.

242 Results also shown that girls tended to change to unhealthier clusters, while boys tended to remain in HC
243 or change to healthier clusters during the transition from childhood to adolescence. Previous literature has
244 shown evidence of sex differences in lifestyle behavioural patterns over the time. Research has shown
245 that girls are more likely to become unhealthier in terms of lifestyle patterns of physical activity and
246 sedentary behaviour [34, 35], especially when they came from lower socio-economic backgrounds [34].
247 Furthermore, a study by Daking et al. [36] found that boys were more likely to show unfavourable
248 changes in lifestyle behavioural clusters (including diet, physical activity, and sedentary behaviour)
249 across the transition from middle to late adolescence compared to girls. These findings show that
250 variability in behaviours included in cluster analyses may influence the direction of sex differences
251 regarding lifestyle patterns. Further research is needed examining the same behavioural clusters over
252 critical time periods to be able to compare and unpack the sex differences in changes over time to inform
253 interventions.

254 In addition to the gender differences seen in changes in clusters over time, indicators of socio-economic
255 status appeared to impact changes in cluster membership. Among both boys and girls, living with a single
256 parent (or other person different from parents) was associated with a higher chance of remaining in UC or
257 changing to unhealthier clusters, while higher parental education was associated with a higher likelihood
258 of remaining in HC. For girls, lower family income was also associated with a higher probability of
259 remaining in UC or changing to unhealthier clusters. Lower socio-economic status has been previously
260 examined in association with lifestyle clusters in young people in cross-sectional studies [9]. For

261 example, low parental education [37–39], low household income [37], and living in a deprived
262 neighbourhood [40] have been associated with unhealthier lifestyle clusters, which may be related to a
263 higher food insecurity/access to healthy foods [41], and/or lower access to a safe built environment that
264 facilitate the physical activity practice. [42, 43]. Contrastingly, we also found that children with low
265 socio-economic status had higher probabilities of transitioning to healthier clusters. These results may be
266 partially explained by our reference group in the multinomial logistic regression analyses which was
267 ‘remaining in HC’. Most of the children from high-socio economic environments were in the healthier
268 cluster at baseline (SF 2), therefore these children had less probability of changing to healthier clusters
269 than children from lower socio-economic groups. The results from this study provide support for
270 intervening on multiple behaviours during childhood and adolescence, and to focus on gender differences
271 and parental characteristics related to socioeconomic disparities. More longitudinal analyses in different
272 countries that help to better understand how young boys and girls from different socioeconomic positions
273 change their lifestyle behaviours over the time may guide research to develop more specific and targeted
274 interventions that better consider the needs of different subgroups of young people.

275 *Strengths and limitations*

276 Strengths of the study include the longitudinal study design, which includes a large nationally
277 representative sample of British children, the inclusion of multiple health behaviours, and thorough
278 analysis of clustering. However, conclusions must be made in light of the limitations. Participants (12%)
279 were excluded from the original sample due to missing data (SF 3), potentially limiting the
280 generalisability of our findings. The outcome variables (physical activity, screen time, and dietetic
281 variables) and sociodemographic variables were all assessed via questionnaires, which could lead to
282 errors in the measurement due to socially desirable responses and/or inaccurate recall [44–46].
283 Furthermore, parents reported during the first wave of data collection and adolescents during the second
284 wave which could change reporting bias [47].

285 **Conclusions**

286 Lifestyle behaviours cluster in children and are susceptible to change over a 7-year period with a high
287 proportion of boys changing to healthier clusters compared to girls, and a higher proportion of girls
288 changing to unhealthier clusters compared to boys. Children from lower socioeconomic status
289 backgrounds including low household income, low parental education or not living with both parents

290 were more likely to change to unhealthier clusters over this period, but also more likely to change to
291 healthier clusters. More longitudinal analyses in different countries are needed for a better understanding
292 of the key drivers of change in clusters among subgroups of children, which is essential for the
293 development of interventions targeting multiple health behaviours.

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440

441 **Statements and declarations**

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449 All authors contributed to the review and editing and the visualization of the paper, as well as all authors
450 read and approved the final draft.

451 **Ethics approval:** All procedures performed for each MCS wave were carried out in accordance with the
452 Declaration of Helsinki and have received ethical approval from the National Research Ethics Service
453 (NRES) Research Ethics Committee (REC) London – Central (REC ref: 13/LO/1786).

454 **Consent to participate:** Parents and participants provided written informed consent prior to completing
 455 all assessments.

456 **Data availability:** Anonymised data from the MCS are accessible for academic use through the UK Data
 457 Service (<https://ukdataservice.ac.uk/>).

458 **Table 1. Baseline characteristics of the sample at age 7 by sex expressed as median (IQR) or n**
 459 **(%).**

	All	Boys	Girls	<i>p</i>
n	9339	4679 (50.1)	4660 (49.9)	
Socio-economic variables				
OECD weekly family income	353.1 (220.2, 530.2)	360.42 (229.5, 533.1)	348.9 (212.5, 527.8)	0.038
OECD < 60% median	2540 (27.2)	1219 (26.0)	1321 (28.3)	0.013
Parental NVQ				0.035
Up to level 3	4468 (47.8)	2240 (47.9)	2228 (47.8)	
Above level 3	3685 (39.5)	1884 (40.3)	1801 (38.6)	
Other qualifications	1186 (12.7)	555 (11.9)	631 (13.5)	
Ethnicity				0.239
White	7810 (83.6)	3934 (84.1)	3876 (83.2)	
Other	1529 (16.4)	745 (15.9)	784 (16.8)	
Household carers				0.231
Both parents	7087 (75.9)	355 (76.0)	3532 (75.8)	
Mother only	1627 (17.4)	794 (17.0)	833 (17.9)	
Father only or other	625 (6.7)	330 (7.0)	295 (6.3)	
Anthropometric variables				

% Body fat	20.0 (17.4, 23.4)	19.0 (16.9, 21.9)	21.2.0 (18.2, 24.7)	<i><0.001</i>
Weight status (IOTF)				<i><0.001</i>
<i>Not overweight</i>	7567 (81.0)	3891 (83.2)	3676 (78.9)	
<i>Overweight</i>	1290 (13.8)	557 (11.9)	733 (15.7)	
<i>Obesity</i>	482 (5.2)	231 (4.9)	251 (5.4)	
Behavioural variables				
MVPA				<i>0.001</i>
<i><5 days per week</i>	9072 (97.1)	4518 (96.6)	4554 (97.7)	
<i>5 or more days per week</i>	267 (2.9)	161 (3.4)	106 (2.3)	
Breakfast frequency				<i><0.001</i>
<i>Not every day</i>	589 (6.3)	239 (5.1)	350 (7.5)	
<i>Every day</i>	8750 (93.7)	4440 (94.9)	4310 (91.5)	
Fruit servings				<i><0.001</i>
<i>< 2 per day</i>	2003 (21.4)	1084 (23.2)	919 (19.7)	
<i>2 or more per day</i>	7336 (78.5)	3595 (76.8)	3741 (80.3)	
TV viewing				<i><0.001</i>
<i>< 3 hours per day</i>	7916 (84.8)	3904 (83.4)	4012 (86.1)	
<i>3 or more hours per day</i>	1423 (15.2)	775 (16.6)	648 (13.9)	
E-games playing				<i><0.001</i>
<i>< 1 hour per day</i>	6079 (65.1)	2677 (57.2)	3402 (73.0)	
<i>1 or more hours per day</i>	3260 (34.9)	2002 (42.8)	1258 (27.0)	

460 a) Significant differences between sexes ($p < 0.05$) are in italics.

461 b) BMI: body mass index; IOTF: International Obesity Task Force; MVPA: moderate-to-vigorous physical
 462 activity; NVQ: National Vocational Qualification; OECD: Organisation for Economic Co-operation and
 463 Development.

464

465 Table 2. Cluster characteristics at baseline and follow-up represented as n (%).

Commented [NP1]: I think this is super useful

Boys								
			Baseline (7 years)			Follow-up (14 years)		
			Healthy Cluster (n = 1945)	Mixed Cluster (n = 1707)	Unhealthy Cluster (n = 1027)	Healthy Cluster (n = 1804)	Mixed Cluster (n = 1440)	Unhealthy Cluster (n = 1435)
MVPA	< 5 days per week		1816 (40.2)	1679 (37.2)	1023 (22.6)	436 (17.5)	819 (32.9)	1237 (49.6)
	5 or more days per week		129 (80.1)	28 (17.4)	4 (2.5)	1368 (62.5)	621 (28.4)	198 (9.1)
Breakfast frequency	Not every day		0 (0.0)	53 (22.2)	186 (77.8)	217 (12.6)	483 (28.1)	1017 (59.2)
	Every day		1945 (43.8)	1654 (37.2)	841 (18.9)	1587 (53.6)	957 (32.3)	418 (14.1)
Fruit servings	< 2 per day		13 (1.2)	418 (38.6)	653 (60.2)	699 (21.3)	1204 (36.8)	1371 (41.9)
	2 or more per day		1932 (53.7)	1289 (35.9)	374 (10.4)	1105 (78.6)	236 (16.8)	64 (4.6)

TV watching	< 3	1945	1503	456 (11.7)	1563	885	333 (12.0)
	hours	(49.8)	(38.5)		(56.2)	(31.8)	
	per day						
E-games playing	3 or	0 (0.0)	204	571 (73.7)	241	555	1102 (58.1)
	more		(26.3)		(12.7)	(29.2)	
	hours						
TV watching	<1 hour	1906	654	117 (4.4)	728	181	38 (4.0)
	per day	(71.2)	(24.4)		(76.9)	(19.1)	
	per day						
E-games playing	1 or	39 (1.9)	1053	910 (45.5)	1076	1259	1397 (37.4)
	more		(52.6)		(28.8)	(33.7)	
	hours						
	per day						
Girls							
Baseline (7 years)				Follow-up (14 years)			
	Healthy Cluster (n = 2453)	Mixed Cluster (n = 1453)	Unhealthy Cluster (n = 754)	Healthy Cluster (n = 885)	Mixed Cluster (n = 1331)	Unhealthy Cluster (n = 2444)	
MVPA	< 5	2351	1449	754 (16.6)	264 (8.0)	924	2095 (63.8)
	days	(51.6)	(31.8)			(28.1)	
	per						
MVPA	5 or	102	4 (3.8)	0 (0.0)	621	407	349 (25.3)
	more	(96.2)			(45.1)	(29.6)	
	days						
	per						

	week							
Breakfast frequency	Not every day	0 (0.0)	129 (36.9)	221 (63.1)	103 (4.1)	560 (22.3)	1848 (73.6)	
	Every day	2453 (56.9)	1324 (30.7)	533 (12.4)	782 (36.4)	771 (35.9)	596 (27.7)	
	< 2 per day	12 (1.3)	463 (50.4)	444 (48.3)	159 (5.0)	641 (20.2)	2375 (74.8)	
Fruit servings	2 or more per day	2441 (65.2)	990 (26.5)	310 (8.3)	726 (48.9)	690 (46.5)	69 (4.6)	
	< 3 hours per day	2450 (61.1)	1244 (31.0)	318 (7.9)	804 (31.6)	861 (33.8)	880 (34.6)	
	3 or more hours per day	3 (0.5)	209 (32.2)	436 (67.3)	81 (3.8)	470 (22.2)	1564 (74.0)	
E-games playing	<1 hour per day	2429 (71.4)	797 (23.4)	176 (5.2)	849 (27.0)	1009 (32.1)	1283 (40.9)	
	1 or more hours per day	24 (1.91)	656 (52.2)	578 (45.9)	36 (2.4)	322 (21.2)	1161 (76.4)	

466 a) Percentages are presented by row.
 467 b) MVPA: moderate-to-vigorous physical activity; TV: television.
 468

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471

472 **Table 3. Changes in behavioural clusters from baseline to follow-up represented as n (%)**

Clusters at 14 y						
Boys (n = 4679)			Girls (n = 4660)			
Clusters at 7 y	Healthy	Mixed	Unhealthy	Healthy	Mixed	Unhealthy
Healthy	893 (19.1)	589 (12.6)	463 (9.9)	609 (13.1)	746 (16.1)	1098 (23.6)
Mixed	649 (13.9)	525 (11.2)	533 (11.4)	207 (4.4)	425 (9.1)	821 (17.6)
Unhealthy	262 (5.6)	326 (7.0)	439 (9.4)	69 (1.5)	160 (3.4)	525 (11.3)

473 a) Percentages have been calculated over the total number of boys and the total number of girls,
 474 respectively.

475

476

477 Table 4. Boys' relative-risk ratios (RRORs) and 95% confidence intervals (CI) for being in the
 478 different longitudinal cluster profiles according to socio economic status and at age 7 (n = 4679).

		Boys							
		Model 1				Model 2			
	<u>Remain</u> <u>ed in</u> <u>Mixed</u> <u>Cluster</u>	<u>Remain</u> <u>ed in</u> <u>Unhealth</u> <u>hy</u> <u>Cluster</u>	<u>Change</u> <u>to an</u> <u>unhealth</u> <u>ier</u> <u>cluster</u>	<u>Chang</u> <u>e to a</u> <u>healthi</u> <u>er</u> <u>cluster</u>	<u>Remain</u> <u>ed in</u> <u>Mixed</u> <u>Cluster</u>	<u>Remain</u> <u>ed in</u> <u>Unhealth</u> <u>hy</u> <u>Cluster</u>	<u>Change</u> <u>to an</u> <u>unhealth</u> <u>ier</u> <u>cluster</u>	<u>Chang</u> <u>e to a</u> <u>healthi</u> <u>er</u> <u>cluster</u>	
OECD <	1.83***	3.04***	1.78***	2.30**	1.21	1.26	1.19	1.43**	
60%	(1.41,	(2.33,	(1.44,	(1.85,	(0.88,	(0.92,	(0.93,	(1.10,	
median	2.39)	3.95)	2.20)	2.85)	1.65)	1.73)	1.53)	1.84)	
NVQ									
Above	0.52***	0.24 ***	0.53***	0.48**	0.58***	0.29***	0.58***	0.54**	
level 3	(0.42,	(0.18,	(0.44,	(0.44,	(0.46,	(0.22,	(0.49,	(0.45,	
	0.66)	0.32)	0.63)	0.63)	0.74)	0.38)	0.70)	0.65)	
Other	1.27	2.10***	1.29	1.87**	1.12	1.95**	1.17	1.51*	
qualificati	(0.85,	(1.44,	(0.929,	(0.40,	(0.72,	(1.30,	(0.82,	(1.63,	
ons	1.91)	3.05)	1.80)	0.57)	1.72)	2.91)	1.66)	2.13)	
Ethnicity									
White	0.79	0.87	0.84	0.64**	0.90	1.24	0.97	0.84	
	(0.58,	(0.62,	(0.66,	*	(0.65,	(0.87,	(0.75,	(0.65,	
	1.07)	1.20)	1.06)	(0.50,	1.25)	1.77)	1.25)	1.09)	

				0.81)				
Household carers								
				1.74**				
Mother only	1.96*** (1.44, 2.67)	3.61*** (2.67, 4.88)	1.87*** (1.46, 2.40)	* (1.34, 2.26)	1.64** (1.18, 2.29)	2.55*** (1.84, 3.55)	1.57** (1.20, 2.05)	1.33* (1.00, 1.76)
Father only or other	2.23** (1.40, 3.55)	3.86*** (2.45, 6.06)	2.06*** (1.40, 3.03)	1.97** (1.32, 2.94)	1.95** (1.22, 3.12)	2.79*** (1.75, 4.43)	1.79** (1.21, 2.64)	1.64* (1.09, 2.47)

479 a) * $p < 0.05$; ** $p < 0.01$; *** $p < 0.001$

480 b) Base values for multinomial logistic regressions were 'remained in healthy cluster' (cluster profile), 'not in
481 poverty: OECD income above 60% of the median' (poverty situation), 'up to level 3' (NVQ), 'other different
482 from white' (ethnicity), and 'both parents' (household carers).

483 c) Model 1: unadjusted; Model 2: adjusted by weight status, % body fat, and the other socio-demographic
484 variables.

485 d) NVQ: National Vocational Qualification.

486

487 Table 5. Girls' **relative-riskodds** ratios (**RRORs**) and 95% confidence intervals (CI) for being in the
 488 different longitudinal cluster profiles according to socio economic status at age 7 (**n = 4660**).

		Girls							
		Model 1				Model 2			
		<u>Remain</u>	<u>Change</u>	<u>Chang</u>	<u>Remain</u>	<u>Change</u>	<u>Chang</u>	<u>Chang</u>	
		<u>ed in</u>	<u>to an</u>	<u>e to a</u>	<u>ed in</u>	<u>to an</u>	<u>e to a</u>	<u>e to a</u>	
		<u>Unhealt</u>	<u>unhealth</u>	<u>healthi</u>	<u>Unhealt</u>	<u>unhealth</u>	<u>healthi</u>	<u>healthi</u>	
		<u>hy</u>	<u>ier</u>	<u>er</u>	<u>hy</u>	<u>ier</u>	<u>er</u>	<u>er</u>	
		<u>Cluster</u>	<u>cluster</u>	<u>cluster</u>	<u>Cluster</u>	<u>cluster</u>	<u>cluster</u>	<u>cluster</u>	
OECD <		2.45***	4.74***	2.65***	2.86	1.07	1.82**	1.38*	1.32
60%		(1.78,	(3.54,	(2.07,	(2.10,	(0.74,	(1.29,	(1.04,	(0.91,
median		3.36)	6.34)	3.40)	3.90)	1.54)	2.56)	1.85)	1.91)
NVQ									
					0.41**				0.44**
Above		0.29***	0.21***	0.39***	*	0.31***	0.27***	0.46***	*
level 3		(0.22,	(0.16,	(0.33,	(0.31,	(0.23,	(0.20-	(0.38,	(0.33,
		0.38)	0.28)	0.48)	0.53)	0.41)	0.35)	0.56)	0.58)
Other		2.10**	3.00***	1.76**	2.96**	1.62	2.25**	1.40	2.14**
qualificati		(1.30,	(1.92,	(1.16,	(1.85,	(0.98,	(1.40,	(0.91,	(1.30,
ons		3.39)	4.70)	2.66)	4.75)	2.69)	3.62)	2.16)	3.52)
Ethnicity									
					0.37**				0.54**
White		0.41***	0.52***	0.60***	*	0.51***	0.88	0.77	0.54**
		(0.29,	(0.37,	(0.46,	(0.26,	(0.35,	(0.61,	(0.57,	(0.37,
		0.57)	7.26)	0.80)	0.52)	0.74)	1.28)	1.03)	0.78)

Househol**d carers**

Mother only	1.93** (1.31, 2.84)	3.69*** (2.62, 5.20)	2.60*** (1.94, 3.50)	1.90** (1.30, 2.78)	1.48 (0.98, 2.23)	2.16*** (1.49, 3.15)	1.93*** (1.41, 2.65)	1.37 (0.91, 2.06)
Father only or other	2.08* (1.14, 3.79)	2.62** (1.49, 4.61)	2.76*** (1.72, 4.41)	0.98 (0.48, 1.99)	1.69 (0.92, 3.10)	1.82* (1.02, 3.25)	2.25** (1.38, 3.58)	0.82 (0.40, 1.68)

489 a) * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

490 b) Base values for the multinomial logistic regressions were 'remained in healthy cluster' (cluster profile), 'not
491 in poverty: OECD income above 60% of the median' (poverty situation), 'up to level 3' (NVQ), 'other
492 different from white' (ethnicity), and 'both parents' (household carers).

493 c) Model 1: unadjusted; Model 2: adjusted by weight status, % body fat, and the other socio-demographic
494 variables.

495 d) NVQ: National Vocational Qualification.

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