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Do neighbourhood characteristics act together to influence BMI? A cross-sectional study of urban parks and takeaway/fast-food stores as modifiers of the effect of physical activity facilities

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Do neighbourhood characteristics act together to influence BMI? A cross-sectional study of urban parks and takeaway/fast-food stores as modifiers of the effect of physical activity facilities

Keywords: built environment; neighbourhoods; physical activity; body mass index; United Kingdom; effect modification; food environment; greenspace

1 ABSTRACT

2 Studies exploring associations between neighbourhood environment and obesity often
3 overlook the fact that neighbourhoods are multi-dimensional and that the effects of one
4 environmental exposure may be modified by another. We examine whether associations
5 between neighbourhood density of formal physical activity (PA) facilities and body mass index
6 (BMI) are modified by the density of neighbourhood green spaces and takeaway stores. We
7 used cross-sectional data from the UK Biobank cohort and linked UK Biobank Urban
8 Morphometric Platform (UKBUMP) for 345 269 urban-dwelling adults aged 40-69. We
9 examined associations between objectively measured BMI and the number of formal PA
10 facilities (gyms, pools, etc.) within 1km of each individual's home, testing separately for
11 interactions with the number of local public green spaces, and number of takeaway stores,
12 within the same 1km buffers. We estimated modifier-stratified associations using
13 multivariable, multilevel models to account for a clustered sampling design and potential
14 confounding. Likelihood ratio tests were used to assess statistical interaction. We found that
15 the association between a greater number of local PA facilities and lower BMI was stronger
16 among people with fewer urban green spaces in their neighbourhood than among those with
17 more green spaces ($P_{\text{interaction}}=0.021$). The same relationship between PA facilities and BMI was
18 also noticeably attenuated among those with more takeaway stores near home, compared with
19 people with none ($P_{\text{interaction}}=0.014$). We conclude that formal PA facilities may buffer against a
20 lack of informal, green resources for PA in areas where the latter are scarce. However, the
21 potential benefits of formal PA facilities in terms of obesity risk may be undermined by an
22 unhealthy food environment close to home. Locating formal PA facilities in places with fewer
23 public green resources and reducing the prevalence of takeaway stores in areas with formal PA
24 resources may maximise the health benefits to be derived from these neighbourhood
25 resources.

26 BACKGROUND

27 Characteristics of neighbourhood environments, such as access to physical activity facilities,
28 green space and takeaway or fast-food outlets, may be linked to obesity risk. However, the
29 evidence base remains inconsistent for many of these neighbourhood exposures (Cobb et al.,
30 2015; Lachowycz and Jones, 2011; Mackenbach et al., 2014; Mayne et al., 2015). One possible
31 explanation for the inconsistencies across studies is that the effects of specific neighbourhood
32 environmental risks may not be universal, but instead vary according to other neighbourhood
33 factors. For example, formal physical activity (PA) facilities are a potentially health-promoting
34 neighbourhood resource (Sallis et al., 2012). Such facilities – e.g. gyms, swimming pools, sports
35 fields – may play a larger role in areas with fewer informal resources that encourage PA (e.g.
36 parks and other urban green space). Conversely, the potential health-promoting influence of
37 the neighbourhood physical activity environment on energy balance and resulting adiposity
38 may be dampened or overridden by the potentially ‘obesogenic’ influence of a neighbourhood
39 food environment dominated by takeaway/fast-food stores. Put another way, the local
40 availability of takeaway stores and of spaces such as parks may act as effect-measure modifiers
41 of the relationship between the formal PA environment and obesity.

42 There is growing recognition that recent increases in obesity prevalence can be viewed as an
43 emergent property of a complex system (Butland et al., 2007; Rutter et al., 2017; Swinburn et
44 al., 2011), and it is therefore important to consider any given exposure or risk factor for obesity
45 within its wider context. The presence of effect modification between neighbourhood
46 characteristics is an example of where context might matter – ignoring the underlying
47 distribution of other effect-modifying neighbourhood characteristics may obscure important
48 effects in some places, and give rise to heterogeneity in findings across different settings. This
49 need to take context into account has been highlighted in a number of recent publications
50 with respect to population health (Craig et al., 2018) and the determinants of the major

51 behavioural risks to health (Panter et al., 2017; Watts et al., 2011). Most studies of
52 neighbourhood built environments and obesity-related outcomes have tended to focus on the
53 influence of single neighbourhood exposures (Feng et al., 2010; Caspi et al., 2012; Van Holle et
54 al., 2012). Yet, recognising the importance of context and the complexity of obesity and its
55 determinants, it follows that attempts to isolate effects of individual neighbourhood
56 characteristics on health can only ever paint an incomplete picture of how environmental
57 factors influence the health of local residents (Giskes et al., 2011; Myers et al., 2016).

58 While there have been studies in the past decade seeking to unpack some of this complexity in
59 various ways, very few have explicitly examined how multiple dimensions of the
60 neighbourhood built environment interact with one another such that one neighbourhood
61 characteristic may modify or moderate the effect of another (Boone-Heinonen et al., 2013).
62 Instead, efforts have been focussed on characterising overall neighbourhood 'obesogenicity' by
63 combining multiple neighbourhood attributes into a single composite measure (Saelens et al.,
64 2012; Tseng et al., 2014) or using methods such as cluster analysis to identify neighbourhood
65 typologies (Adams et al., 2013; Hobbs et al., 2018a; Meyer et al., 2015). By understanding
66 whether the effect of one neighbourhood characteristic is modified by the presence of other
67 neighbourhood characteristics, we may better describe how neighbourhoods shape health and
68 behaviour. We may also start to identify settings in which interventions targeting a particular
69 feature of the built environment may have greater (or lesser) potential for reducing or
70 preventing obesity in the populations residing there, and optimise future interventions
71 accordingly (Economos et al., 2015).

72 In this paper we focus on potential modification of the relationship between the formal PA
73 environment and adiposity by neighbourhood availability of parks, and neighbourhood food
74 environment. Many formal PA facilities are businesses, and as such they are potentially
75 modifiable via regulatory and commercial levers. Some are run by local authorities, and thus

76 are potentially also amenable to other policy interventions aimed at locating these facilities
77 where they may have the greatest benefit to local populations. In a recent cross-sectional
78 study using UK Biobank, we observed a pattern of lower mean waist circumference, BMI and
79 body fat associated with increasing neighbourhood density of PA facilities (Mason et al., 2018).
80 As described above, other neighbourhood resources such as parks and fast food outlets are
81 potential effect-measure modifiers of these relationships; parks because they can provide
82 alternative opportunities for informal outdoor PA that may be more accessible and appealing
83 than formal PA facilities, and takeaway/fast-food outlets because unhealthy food
84 environments may negate healthy PA environments. Examining these elements of the
85 relationship between the formal PA environment and BMI may deepen our understanding of
86 where intervening on the formal PA environment may be more beneficial, or, alternatively,
87 other modifiable neighbourhood attributes that may boost the potential for local residents to
88 benefit from local formal PA facilities.

89 We assess these possible environmental effect-modification relationships among adults living
90 in urban residential areas in the United Kingdom by testing the following hypotheses. First,
91 that the availability of formal PA facilities will be more strongly associated with BMI among
92 people with no parks or other public open/green spaces near their home than it is among
93 those with greater park availability. Second, that the association between formal PA resources
94 and BMI will be weaker among people with more takeaway stores near their home than it is
95 among those with no or fewer takeaway stores near home.

96

97 **METHODS**

98 *Study design and data collection*

99 In this cross-sectional study, we used baseline data from the UK Biobank (project 17380), a

100 population-based sample of half a million UK adults for whom the scientific rationale and
101 more detailed study design and survey methods of UK Biobank have been described elsewhere
102 (UK Biobank, 2007). Data were potentially available from 502, 656 individuals who visited one
103 of 22 UK Biobank assessment centres across the United Kingdom between 2006 and 2010,
104 where they provided baseline data spanning health, sociodemographics, behaviour,
105 psychosocial and other factors, and were subject to a range of sampling and testing
106 procedures. All individuals aged 40–69 years living within a 25-mile radius of an assessment
107 centre and listed on National Health Service (NHS) patient registers were invited to
108 participate in the study. The age range was chosen by UK Biobank as an important period for
109 the development of many chronic diseases.

110 *Local environment data*

111 Linked to UK Biobank is the UK Biobank Urban Morphometric Platform (UKBUMP), a high-
112 resolution spatial database of objectively measured characteristics of the physical environment
113 surrounding each participant's exact residential address. Environmental data in UKBUMP are
114 derived from multiple national spatial datasets using automated processes (Sarkar et al., 2015).
115 The available measures of the local environment include: densities of various land uses;
116 proximity to various health-relevant destinations (e.g. GP practices, industrial sites, fast-food
117 outlets); street network accessibility; and pollution. The metrics were constructed using data
118 collected during the baseline individual assessment phase. No environmental data were
119 collected for the Stockport assessment area, which was the UK Biobank pilot site, leaving 21
120 assessment areas in scope.

121 *Outcomes*

122 Our primary outcome of interest was Body Mass Index (BMI, kg/m²). Height and weight
123 measurements were made by trained staff using standard procedures (UK Biobank, 2007) and

124 BMI was treated as a continuous variable. In sensitivity analyses, waist circumference (in cm,
125 measured manually by trained nurses) and body fat percentage (measured using a
126 bioimpedance machine) were examined as secondary outcomes to assess the consistency of
127 the results across alternative measures of adiposity.

128 *Primary exposure*

129 Our primary exposure was the total number of formal PA facilities within a one-kilometre
130 street-network buffer around each individual's place of residence (categorised as 0, 1, 2-3, 4 or
131 more, to account for the data being positively skewed). Formal PA facilities were defined at
132 address level as any land use classified in the Commercial-Leisure subcategory of the UK
133 Ordnance Survey AddressBase Premium database. This subcategory comprises a range of
134 indoor and outdoor facilities designed for sporting and leisure activities, such as gyms,
135 swimming pools and playing fields (for details see Supplementary Material). A 1-km buffer has
136 been used in numerous other studies; it equates to about a 10-15 minute walk and has been
137 reported to be roughly the area that people perceive to be their neighbourhood (Lee and
138 Moudon, 2006).

139 *Potential effect-measure modifiers*

140 To test hypothesis 1 (i.e. that the association between availability of PA facilities and BMI will
141 be stronger among people with no parks near their home), we examined effect-measure
142 modification by *urban park availability*, measured as the number of parks or other public
143 open/green spaces in a one-kilometre street-network buffer around each participant's home
144 address. As with formal PA facilities, the number of parks is derived from the UK Ordnance
145 Survey AddressBase Premium database, and we included any land use categorised as Park;
146 Public Park/Garden; Public Open Space/Nature Reserve; Open Space/Heath/Moorland; or
147 Playground. The distribution of the number of these sites in a buffer was highly positively

148 skewed, and for this analysis was therefore categorised as no parks, one or two parks, or at
149 least three parks.

150 To test hypothesis 2 (i.e. that the association between PA facilities and BMI will be weaker
151 among people with more takeaway stores near their home), we examined possible effect-
152 measure modification by *takeaway/fast-food store availability*, measured as the number of
153 addresses classified as a 'hot/cold fast-food outlet/takeaway' in the UK Ordnance Survey
154 AddressBase Premium database (Sarkar et al., 2015), again within a one-kilometre street-
155 network buffer around each participant's home address. As with parks, the distribution of the
156 number of takeaways in a buffer was highly skewed, so we categorised the count into three
157 levels (0 / 1-2 / 3 or more).

158 *Statistical analysis*

159 We first examined the distributions of the various neighbourhood attributes across the sample
160 by cross-tabulating categories of the number of formal PA facilities in people's
161 neighbourhoods with categories of each of the potential modifiers (urban park availability and
162 takeaway store availability). To test each effect-measure modification hypothesis, we
163 compared multilevel linear models of the independent association between the formal PA
164 environment and BMI with and without interaction terms for the product of the formal PA
165 environment and each potential modifier. Multilevel models were used to account for the
166 clustering by assessment centre in the sampling design, and were estimated with random
167 intercepts and random coefficients for the main exposure. We used likelihood ratio tests to
168 compare the models and we report the P value from these tests to indicate the strength of the
169 evidence against the null hypothesis of no effect modification on the additive scale. We then
170 stratified the sample by the potential effect modifier and estimated stratum-specific mean
171 differences (and 95% CIs) in BMI for categories of increasing numbers of formal PA facilities,

172 relative to people with no PA facilities with one kilometre of home.

173 Models were adjusted for potential confounding by age (years), sex (male or female), self-
174 reported ethnicity (white, south Asian, black, other Asian, mixed white and black, mixed
175 white and Asian, mixed other, or other), highest education level attained (college or university
176 degree, post compulsory education, higher secondary education, secondary education,
177 vocational qualifications, other professional qualification, or none of the above), annual
178 household income (<£18 000, £18 000–30 999, £31 000–51 999, £52 000–100,000, or >£100 000),
179 employment status (paid work, retired, unable to work, unemployed, or other), area
180 deprivation (2001 Townsend index score (National Centre for Research Methods); higher
181 score=greater deprivation), and neighbourhood residential density (count of residential
182 dwellings within a 1-km street-network buffer of home address, log transformed). The focus of
183 the analysis was to identify moderation or enhancement of the estimated effect of the primary
184 exposure – the formal PA environment – by other neighbourhood features (specifically
185 parks/fast-food). Therefore, we adjusted each model for the set of covariates that represents
186 potential confounders of the relationship between the primary exposure and the outcome, and
187 which we identified with the aid of a directed acyclic graph. While adjustment for the
188 confounders of the modifier-outcome associations was therefore not essential (Vanderweele,
189 2015, p. 269), those sets of confounders are, in this context, likely to be very similar.

190 We also adjusted each model for the other potential modifier not under examination in that
191 model (i.e. we included takeaway store availability as a covariate in the models testing for
192 effect modification by number of parks, and controlled for urban park availability in the model
193 testing for effect modification by takeaway store availability). This made no substantive
194 difference to the point estimates but slightly improved precision of the estimates. In a
195 previous analysis of the association between the formal PA environment and adiposity, we also
196 found that adjustment for diet (total energy intake) did not lead to substantively different

197 conclusions, but did artificially inflate point estimates due to extensive missing data;
198 therefore, we deemed it inappropriate to adjust the models in this paper for diet, particularly
199 as we identified an additional risk of inducing collider bias by adjusting for diet (Mason et al.,
200 2018).

201 *Missing data and sample restrictions*

202 Perceptions of access to food outlets and public amenity of parks and other public open/green
203 spaces are both likely to differ in urban residential areas compared with non-urban areas
204 (Dennis and James, 2017; Maas et al., 2006). For example, to a person living in a rural area,
205 many facilities will be relatively far away, so having takeaway stores near home won't mean
206 the same thing as it does to someone living in an urban area if both are measured on the same
207 scale. And in rural areas close to natural landscapes, parks in the immediate neighbourhood
208 may be less important as a potential site of PA than they are for people in the middle of a city.
209 We therefore restricted the analysis to the 86% of the UK Biobank cohort living in areas that
210 are classified by the Office of National Statistics as urban (specifically, where a person's home
211 postcode is located within a city or a town with a population of at least 10,000 people).

212 Participants without complete data on all covariates were excluded from analysis.
213 Approximately three percent of individuals were missing data on their neighbourhood
214 environment. Data for all other variables were missing at a frequency of <2%, with the
215 exception of income (14.9% missing). The final complete case sample comprised 345, 269
216 individuals.

217 *Sensitivity analyses*

218 Although BMI is a very widely used measure of adiposity, other objective measures may be
219 better predictors of adiposity-related ill-health in some segments of the population, and all
220 measures have their own strengths and weaknesses. We checked for consistency of our

221 findings across alternative measures of adiposity, by repeating the analyses for waist
222 circumference and body fat percentage rather than BMI. To examine the impact of our
223 decision to exclude non-urban participants rather than adjust models for urbanicity, we
224 repeated the primary analyses on the full urban and non-urban sample combined, adjusting
225 for urban/non-urban status. A proximity measure of takeaway stores (distance from home
226 address to nearest store) was also available in the dataset so we also checked whether our
227 results were sensitive to the measure of the fast-food environment. Although density and
228 proximity measures of the UK food environment have been shown to be correlated, they are
229 nonetheless theoretically distinct constructs (Burgoine et al., 2013).

230 *Ethics approval*

231 As part of an approved UK Biobank research project, the study is covered by the ethics
232 approval granted to UK Biobank by the North West Multi-centre Research Ethics Committee
233 (reference number 16/NW/0274). Institutional ethics approval was also granted via the
234 London School of Hygiene and Tropical Medicine's Observational/Intervention Research
235 Ethics Committee in September 2016 (LSHTM Ethics Reference 11897).

236

237 **RESULTS**

238 The mean BMI of the analytical sample was 27.5kg/m², and the median number of formal PA
239 facilities within a 1000m street-network distance of participants' homes was two, with just
240 over a quarter of participants having no facilities close to home (Table 1). Participants had a
241 median of one park or other public green space within one kilometre of home, and 41% had no
242 parks near home. Half the sample had no takeaway stores within one kilometre of home, while
243 23% of the sample lived within a kilometre of three or more such stores.

244 [Insert Table 1 here]

245 There is a strong positive correlation between number of parks and number of formal PA
246 facilities, with 58% of people who have no formal PA facilities within 1km of home also having
247 no parks close to home, while 41% of those with at least four PA facilities nearby also have at
248 least three parks nearby (Table 2). Number of takeaway/fast-food stores is also correlated with
249 access to formal PA facilities, reflecting the clustering of commercial and public services in
250 more densely populated areas. Seventy percent of people with no nearby PA facilities also have
251 no takeaway stores within 1km of their home, compared to only one third of people with at
252 least four PA facilities having no takeaways near home.

253 [Insert Table 2 here]

254

255 Comparison of models with and without a product term for the interaction between formal PA
256 facilities and parks suggests the number of parks within one kilometre of a person's home
257 does modify the association between the formal PA environment and BMI ($P_{\text{interaction}}=0.021$).
258 Figure 1 shows estimates of the association between formal PA facilities and BMI, stratified by
259 urban park availability. As hypothesised, stratification shows that the association between PA
260 facilities and BMI is weak in the areas with the most urban parks, while in contrast there is a
261 clear association between lower BMI and higher density of formal PA facilities in the areas
262 with no parks (Figure 1). People living in areas with no parks have a mean BMI 0.21 kg/m²
263 smaller if they have 2-3 formal PA facilities near home, and 0.48kg/m² smaller if they have at
264 least four PA facilities near home, when compared with those who have neither PA facilities
265 nor parks within a kilometre of home.

266 [Insert Figure 1 here]

267 There is also good statistical evidence that the PA environment-BMI association is modified by
268 availability of takeaway/fast-food stores ($P_{\text{interaction}}=0.014$). In line with our second hypothesis,
269 stratified results showed that among people living in areas with at least three takeaway stores,
270 the association between density of nearby formal PA facilities and BMI is considerably weaker
271 than it is among those who live in areas with fewer or no takeaway stores (Figure 2). Among
272 people with three or more takeaways near home, those with at least four PA facilities near
273 home have a mean BMI 0.22 kg/m² smaller than those with no PA facilities within a kilometre
274 of home, but among people with none, one or two takeaways near home, the magnitude of
275 that effect is more than doubled.

276 [Insert Figure 2 here]

277 Results were broadly consistent across alternative measures of adiposity: very similar patterns
278 of effect-measure modification to those we observed for BMI were also present for waist
279 circumference and particularly body fat percentage (see Supplementary Material). When we
280 included respondents living in non-urban areas and adjusted for urban/non-urban status, the
281 patterns across stratum-specific models mirrored those observed in the urban-only sample.
282 Statistical evidence of an interaction was similar (for park availability) or stronger (for
283 takeaway store availability) (see Supplementary Material). The primary results were also
284 replicated when using proximity to takeaway store as an alternative measure of the fast-food
285 environment – in that case the primary association between PA environment and BMI was
286 notably weaker among people living within 500m of a store compared with those living further
287 away ($P_{\text{interaction}} < 0.001$).

288

289 **DISCUSSION**

290 In this study we examined whether the association between neighbourhood availability of
291 formal PA facilities and BMI is moderated by the availability of parks and takeaway food stores
292 in the neighbourhood environment. We found evidence to support the presence of
293 environmental effect-measure modification in this large sample of mid-life adults from across
294 the UK. In stratified models we observed that the relationship between greater access to
295 formal PA facilities and lower BMI is stronger among people living in areas with fewer urban
296 parks and other public open/green spaces than it is among people with more of these, where
297 there is likely to be greater opportunity for informal, outdoor PA. Furthermore, the association
298 between the formal PA environment and BMI is of a much smaller magnitude among people
299 living close to several takeaway stores than it is among those who do live near fewer or no
300 such stores.

301 These findings suggest that locating formal PA facilities close to residential areas has the
302 potential to reduce BMI among local residents, but that other contextual features of the
303 neighbourhood are likely to influence these potential benefits. To date, the formal PA
304 environment has received less research attention than some other neighbourhood
305 characteristics, particularly in Europe, and findings have been inconsistent (Mackenbach et
306 al., 2014; Sallis et al., 2012). Our deeper examination of its relationship with BMI sheds new
307 light on settings where intervening on the formal PA environment may be more beneficial,
308 and how other neighbourhood attributes could be modified to boost the potential for local
309 residents to benefit from local formal PA facilities.

310 While we cannot infer causality from this cross-sectional observational study, our results
311 suggest that increasing the availability of formal PA facilities may have the most potential to
312 reduce population obesity in areas that have the lowest densities of parks and least exposure
313 to takeaway/fast-food stores. This is consistent with the hypothesis that in areas with fewer
314 parks and other green spaces, formal PA facilities provide valuable opportunities for PA that

315 are otherwise lacking. Meanwhile, formal PA facilities may have limited influence in areas
316 with takeaway stores close to people's homes – even if they do serve to increase PA there, our
317 findings suggest the positive benefits for body weight may be dampened by the influence of an
318 unhealthy food environment. As a public health intervention, the introduction of PA resources
319 such as gyms, swimming pools and other sports facilities in neighbourhoods with numerous
320 takeaway stores may be ineffective unless coupled with other interventions aimed at
321 minimising the influence of those stores. In urban areas well served by parks, interventions
322 involving formal PA facilities may not be a priority and a focus on other environmental
323 interventions may be more effective in improving population health. More causally focused
324 study designs are needed to confirm these implications.

325 Our findings also highlight the possibility that closer examination of the relationships
326 between other neighbourhood exposures and obesity-related outcomes may also reveal effect
327 heterogeneity according to additional characteristics of the area. Such heterogeneity may, at
328 least partially, explain inconsistent results across studies and settings. We examined only
329 three neighbourhood characteristics, motivated by two plausible effect modification
330 hypotheses. Similar interactions may also exist between other neighbourhood characteristics.

331 PA environment effects on obesity in the UK have sometimes been shown to be concentrated
332 in individuals with higher incomes (Mason et al., 2018) or more education (Hobbs et al.,
333 2018b). While we have not examined an additional interplay with socioeconomic status in this
334 paper, it is likely that an important caveat applies to our findings: that the potential for the
335 benefits of local access to formal PA facilities to be maximised via a supportive broader
336 neighbourhood environment relies on PA facilities being accessible and affordable for all.

337 *Strengths & Limitations*

338 These findings are a novel contribution to this area of research. To our knowledge, no other
339 studies in the UK, and few outside it, have explicitly examined modification of the association
340 between the formal neighbourhood PA environment and adiposity (or any other obesity-
341 related outcome) by other neighbourhood built environment characteristics. One similar
342 study in the United States concluded that combined changes to the food and PA
343 environments would have stronger and more consistent effects on BMI than changes that
344 addressed only one dimension or the other. Our findings provide similar evidence in a
345 European context (Boone-Heinonen et al., 2013). Others have examined composite measures
346 of neighbourhood obesogenicity or other similar constructs, and while such research
347 importantly recognises and draws attention to the complex and multidimensional nature of
348 neighbourhood environments, it lends itself to more general conclusions about the
349 importance of holistic healthy urban planning, rather than moving towards specific policy
350 recommendations. Furthermore, these studies typically rely on data-driven approaches such
351 as latent class analysis, and this makes generalising to other populations challenging. Here,
352 although our findings require confirmation using longitudinal data and more causally
353 focussed methods, the results provide evidence in support of two clearly defined and
354 theoretically grounded effect modification hypotheses, and point to prioritisation of built
355 environment interventions that take into account local context. Our findings were also
356 consistent across multiple adiposity measures, and robust to various modelling choices
357 including operationalisation of the effect modifiers, as shown in the sensitivity analyses we
358 performed.

359 UK Biobank is a very large and geographically diverse cohort, providing unique opportunities
360 in this field of research. However, as a sample it may not be representative of the broader
361 population, being based on only a 5.5% response rate, and indeed, the sample does show some
362 evidence of 'healthy volunteer' bias (Fry et al., 2017). A further potential source of selection

363 bias is the exclusion of a large number of observations without data on household income.
364 With its focus on adults in mid-life, we also cannot generalise our conclusions beyond this age
365 range. Due to the cross-sectional nature of this study it is impossible to draw strictly causal
366 inferences about these patterns of association, but the results do lend support to two *a priori*
367 hypotheses about plausible interactive effects of multiple aspects of the neighbourhood
368 environment. There remains the possibility that the observed main associations are driven by
369 people with lower BMIs self-selecting into 'healthier' neighbourhoods. Studies that have
370 directly examined the influence of self-selection on neighbourhood-health effects have
371 reached inconsistent conclusions about the likely bias this may induce (Grafova et al., 2014;
372 James et al., 2015; McCormack and Shiell, 2011). For those fortunate enough to have substantial
373 choice over where they live, the presence of formal PA facilities alone is unlikely to be a major
374 governing factor in that choice, but the presence of such facilities is likely to coincide with
375 other facilities that may enhance the desirability of a neighbourhood, including parks and
376 other green space (as we observed in Table 2). That said, in this sample, the individuals living
377 in neighbourhoods with high densities of parks and PA facilities are, contrary to expectation,
378 not necessarily those with the highest incomes, or living in the least deprived postcodes in the
379 study.

380 Due to the size of the sample and the breadth of the neighbourhood characterisation, large-
381 scale automated processes were used to derive the environmental metrics on which we have
382 relied here (Sarkar et al., 2015). While the best readily available for conducting these analyses
383 at scale, those metrics are of varying quality, accuracy and suitability. It should be noted that
384 the takeaway store measure in particular may be susceptible to some misclassification error:
385 the source database is supplied by local authorities and might misclassify some fast-food
386 outlets as restaurants rather than takeaways. In addition, the measure of park availability does
387 not account for the quality of those spaces, and we also do not have any information about

388 whether respondents actually use their local PA facilities. There is also a possibility that if any
389 of the main associations are confounded in one stratum of the potential effect modifier and
390 not another, we may erroneously infer effect modification when none is present (Kamangar,
391 2012).

392 These findings therefore provide preliminary observational evidence for plausible interplay
393 between multiple aspects of the built environment in the UK, but further research using more
394 causally focussed approaches such as longitudinal or quasi-experimental study designs is
395 needed. An additional implication of our findings is that evaluations of PA environment
396 interventions, particularly those pertaining to formal PA facilities, may be underestimating the
397 impact of the intervention if possible moderation by local park availability and food
398 environments are ignored.

399 **CONCLUSIONS**

400 Residential neighbourhoods are complex and multidimensional and the examination of the
401 effect of individual environmental characteristics on obesity in isolation overlooks this
402 complexity. Here we examined whether some neighbourhood characteristics modify the effect
403 of others to better understand how they may operate in concert to influence BMI. Our
404 findings suggest that formal PA facilities may buffer against a lack of informal, green resources
405 for PA such as parks, in areas where the latter are scarce, but that potential benefits of formal
406 PA facilities in terms of BMI may be undermined by the presence of takeaway/fast-food stores
407 close to home. Reducing the prevalence of unhealthy food stores in areas with formal PA
408 resources, and prioritising the location of formal PA facilities in places without public parks,
409 may maximise the potential for PA facilities to influence BMI. An approach to urban planning
410 that takes into account moderating effects of other neighbourhood characteristics is required
411 in order to maximise the population health benefits of the urban built environment.

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Table 1. Summary of sample characteristics (N=345 269)

<i>BMI (kg/m²)</i>	Mean (SD)	27.5 (4.8)
Formal PA environment		
<i>Number of facilities in 1km buffer</i>	Median (IQR)	2 (0 - 4)
0	n (%)	94 717 (27.4)
1	n (%)	67 570 (19.6)
2-3	n (%)	86 006 (24.9)
4 or more	n (%)	96 976 (28.1)
Park availability		
<i>Number of parks and other public/open green spaces in 1km buffer</i>	Median (IQR)	1 (0 - 3)
0	n (%)	142 491 (41.3)
1 or 2	n (%)	108 803 (31.5)
3 or more	n (%)	93 975 (27.2)
Fast food environment		
<i>Number of takeaway/fast-food stores in 1km buffer</i>	Median (IQR)	0 (0 - 2)
0	n (%)	174 405 (50.5)
1 or 2	n (%)	91 030 (26.4)
3 or more	n (%)	79 834 (23.1)
Covariates		
<i>Age</i>	Mean (SD)	56.1 (8.1)
<i>Sex (female)</i>	n (%)	181 896 (52.7)
<i>Ethnicity</i>		
White	n (%)	327 583 (94.9)
South Asian/South Asian British	n (%)	5 295 (1.5)
Black/Black British	n (%)	5 323 (1.5)
Chinese/other(non-South) Asian	n (%)	2 209 (0.6)
Mixed: White/Black	n (%)	733 (0.2)
Mixed: White/Asian	n (%)	598 (0.2)
Mixed: Detail unknown	n (%)	709 (0.2)
Other	n (%)	2 819 (0.8)
<i>Income</i>		
Less than 18,000	n (%)	81 544 (23.6)
18,000 to 30,999	n (%)	88 756 (25.7)
31,000 to 51,999	n (%)	89 795 (26.0)
52,000 to 100,000	n (%)	68 051 (19.7)
Greater than 100,000	n (%)	17 123 (5.0)
<i>Education</i>		
College or University degree	n (%)	117 441 (34.0)
A levels/AS levels or equivalent	n (%)	39 526 (11.5)

O levels/GCSEs or equivalent	n (%)	74 022 (21.4)
CSEs or equivalent	n (%)	19 594 (5.7)
NVQ or HND or HNC or equivalent	n (%)	23 404 (6.8)
Other professional qualifications	n (%)	17 390 (5.0)
None of the above	n (%)	53 892 (15.6)
<i>Employment status</i>		
Paid employment or self-employed	n (%)	210 595 (61.0)
Retired	n (%)	106 508 (30.9)
Unable to work	n (%)	10 782 (3.1)
Unemployed	n (%)	5 339 (1.6)
Home duties/carer/student/volunteer/other	n (%)	12 045 (3.5)
<i>Area deprivation (2001 Townsend index)</i>	Median (IQR)	-2.0 (-3.6 - 0.7)
<i>Residential density (residential address points per 1km street network buffer)</i>	Median (IQR)	2 152 (1 352 – 3 344)
<i>Assessment area</i>		
Manchester	n (%)	11 012 (3.2)
Oxford	n (%)	7 328 (2.1)
Cardiff	n (%)	12 736 (3.7)
Glasgow	n (%)	14 656 (4.2)
Edinburgh	n (%)	11 714 (3.4)
Stoke	n (%)	12 283 (3.6)
Reading	n (%)	18 217 (5.3)
Bury	n (%)	20 192 (5.9)
Newcastle	n (%)	26 432 (7.7)
Leeds	n (%)	30 831 (8.9)
Bristol	n (%)	27 796 (8.1)
Central London	n (%)	8 241 (2.4)
Nottingham	n (%)	20 670 (6.0)
Sheffield	n (%)	20 849 (6.0)
Liverpool	n (%)	25 037 (7.3)
Middlesborough	n (%)	12 954 (3.8)
Hounslow	n (%)	21 321 (6.2)
Croydon	n (%)	21 126 (6.1)
Birmingham	n (%)	20 060 (5.8)
Swansea	n (%)	1 503 (0.4)
Wrexham	n (%)	311 (0.1)

Table 2. Bivariate associations between neighbourhood characteristics

Formal PA facilities	<i>Availability of parks and other open/green spaces</i>							
	No parks		One or two parks		≥ 3 parks		Total	
	n	%	n	%	n	%	n	%
0	54,536	57.58	25,196	26.60	14,985	15.82	94,717	100.00
1	31,684	46.89	20,888	30.91	14,998	22.20	67,570	100.00
2-3	32,035	37.25	29,423	34.21	24,548	28.54	86,006	100.00
4 or more	24,236	24.99	33,296	34.33	39,444	40.67	96,976	100.00
<i>Total</i>	<i>142,491</i>	<i>41.27</i>	<i>108,803</i>	<i>31.51</i>	<i>93,975</i>	<i>27.22</i>	<i>345,269</i>	<i>100.00</i>

Formal PA facilities	<i>Availability of takeaway/fast-food stores</i>							
	No stores		One or two stores		≥ 3 stores		Total	
	n	%	n	%	n	%	n	%
0	66,096	69.78	20,964	22.13	7,657	8.08	94,717	100.00
1	37,621	55.68	19,149	28.34	10,800	15.98	67,570	100.00
2-3	38,499	44.76	26,278	30.55	21,229	24.68	86,006	100.00
4 or more	32,189	33.19	24,639	25.41	40,148	41.40	96,976	100.00
<i>Total</i>	<i>174,405</i>	<i>50.51</i>	<i>91,030</i>	<i>26.36</i>	<i>79,834</i>	<i>23.12</i>	<i>345,269</i>	<i>100.00</i>

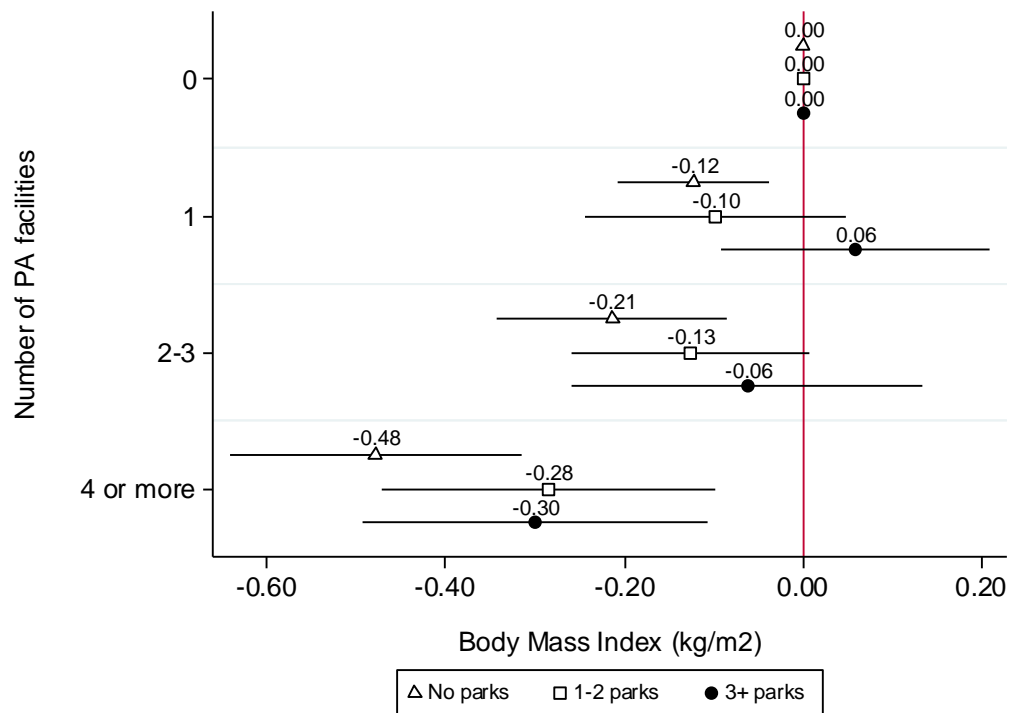


Figure 1. Association between number of formal PA facilities and BMI, stratified by park availability

Figure shows park availability-stratified, fully adjusted mean differences in BMI and associated 95% CIs from multilevel linear regression models. The dashed line at zero represents the reference category (no physical activity facilities with 1km of home). Models are adjusted for age, sex, ethnicity, area deprivation, individual socioeconomic characteristics (income, education, and employment status), residential density, and takeaway store availability.

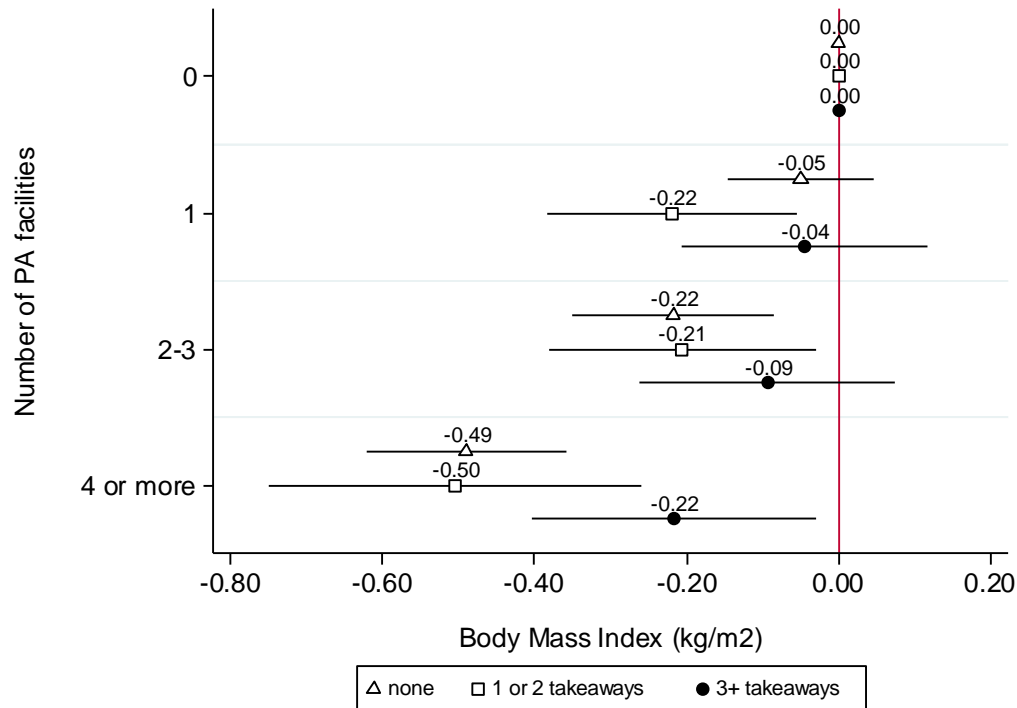


Figure 2. Association between number of formal PA facilities and BMI, stratified by takeaway store availability.

Figure shows takeaway availability-stratified, fully adjusted mean differences in BMI and associated 95% CIs from multilevel linear regression models. The dashed line at zero represents the reference category (no physical activity facilities with 1km of home). Models are adjusted for age, sex, ethnicity, area deprivation, individual socioeconomic characteristics (income, education, and employment status), residential density, and park availability.

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Highlights

- Neighbourhoods are multi-dimensional; dimensions may interact to influence health.
- Stronger association of PA facility availability and lower BMI when no local parks.
- Weaker association between PA facilities and BMI when more takeaways near home.
- Formal PA facilities may buffer against a lack of informal, green spaces for PA.
- Benefits of PA facilities may be undermined by unhealthy local food environments.

KM designed the analysis with input from SC and NP. KM undertook the data management, statistical analysis and writing of the manuscript. SC and NP contributed to the interpretation of results and drafting of the manuscript. All authors approved the final submitted version.

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