

Doses of Neighborhood Nature: Benefits for Mental Health of Living with Nature

Running title: Variation in Nature and Mental Health

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fraction, Urban nature.

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3 22 **Summary**
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5 23 Experiences of nature provide many mental health benefits, particularly for people
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7 24 living in urban areas. The natural characteristics of city residents' neighborhoods are
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9 25 likely to be critical determinants of the daily nature dose that they receive, however
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11 26 which characteristics are important remains unclear. One possibility is that the
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13 27 greatest benefits are provided by characteristics that are most visible during the day
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15 28 and so most likely to be experienced by people. We demonstrate that of five
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17 29 neighborhood nature characteristics tested, vegetation cover and afternoon bird
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19 30 abundances were positively associated with a lower prevalence of depression, anxiety
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21 31 and stress. Further, dose-response modelling shows a threshold response where the
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23 32 population prevalence of mental health issues is significantly lower beyond minimum
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25 33 limits of neighborhood vegetation cover (depression >20% cover, anxiety >30%
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27 34 cover, stress >20% cover). Our findings demonstrate quantifiable associations of
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29 35 mental health with the characteristics of nearby nature that people actually experience.
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36 **Background**

37 The economic costs of anxiety and mood disorders, such as depression, have been
38 estimated at €187.4 billion.yr⁻¹ for Europe alone (Gustavsson et al. 2012, Olesen et al.
39 2012). Alongside stress, they are some of the most prevalent work-related health
40 issues (13.7% of all reported work-related cases; Eurostat 2012). This growing
41 problem has, at least in part, been attributed to the increasing disconnect between
42 people and the natural world that is resulting from more urbanized, sedentary
43 lifestyles (the ‘extinction of experience’; Miller 2005, Soga and Gaston 2015). This is
44 supported by research that shows interactions with nature promote psychological
45 restoration (Kaplan 1995), improved mood (Hartig et al. 2003, Barton and Pretty 2010,
46 Roe and Aspinall 2011), improved attention (Hartig et al. 2003, Ottosson and Grahn
47 2005) and reduced stress and anxiety (Ulrich et al. 1991, Grahn and Stigsdotter 2003,
48 Hartig et al. 2003, Maas et al. 2009).

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50 The causal factors behind poor mental health are complex and diverse (Kinderman et
51 al. 2015), and cultural and socio-economic differences between regions may influence
52 responses to interactions with nature (reviewed by Keniger et al. 2013).

53 Understanding and capitalizing on the mechanisms by which natural environments
54 provide psychological benefits nonetheless has the potential to be a novel and cost-
55 effective approach to reducing the prevalence of some forms of mental ill health
56 (Hartig et al. 2014, Shanahan et al. 2015b). Indeed, nature is likely to influence
57 mental health through a range of mechanistic pathways (Shanahan et al. 2015b).

58 Attention restoration theory proposes that the natural world promotes recovery from
59 mental fatigue that occurs during the performance of cognitive tasks that require
60 prolonged maintenance of directed attention (Kaplan 1995), while stress reduction

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3 61 theory argues that natural environments facilitate reductions in physiological arousal
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5 62 following stress (Ulrich et al. 1991). Both of these complementary theoretical
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7 63 frameworks lead to improved mental health from experiencing nature, through
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10 64 decreased rumination, increased cognition and reduced stress (Berman et al. 2012;
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12 65 Jiang et al. 2014; Tyrväinen et al. 2014; Bratman et al. 2015).

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16 67 Increasingly, evidence suggests that the availability and quality of neighborhood
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18 68 green spaces are associated with greater well-being (White et al. 2013) and lower
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20 69 levels of depression, anxiety and stress (Beyer et al. 2014). These benefits may be
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22 70 gained from intentionally interacting with nature (such as through visiting
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24 71 neighborhood green spaces or spending time in a garden), from incidental interactions
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26 72 whereby people are exposed to nature as they engage in other activities (such as
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28 73 walking to the shops), or indirectly while not actually being present in nature (such as
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30 74 viewing it through a window; Keniger et al. 2013). The natural environment around
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32 75 the home is the nature that most people will experience every day, and therefore
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34 76 through all three kinds of nature interactions will significantly contribute towards
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36 77 people's daily nature experience.

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42 79 To date, most research into the health benefits of nature has considered the role of
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44 80 green spaces per se. The role of specific biological components of those spaces
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46 81 remains unclear, although these need to be identified effectively to guide planning to
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48 82 operationalize the use of nature as a health promoting tool. In urban areas, two of the
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50 83 most visible elements of nature are vegetation cover and bird communities. The
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52 84 presence of vegetation has been found to have positive mental health benefits,
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54 85 including, but not limited to, helping to reduce stress and promoting restoration from
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3 86 mental fatigue (e.g., Fuller et al. 2007, Alvarsson et al. 2010, Dallimer et al. 2012).
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5 87 Having more bird species in the environment and watching birds has been shown to
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7 88 be good for people's psychological well-being (Fuller et al. 2007, Curtin 2009, Brock
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9 89 et al. 2015; Cox and Gaston 2016), while listening to bird song has been shown to
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11 90 contribute towards perceived attention restoration and stress recovery (Ratcliffe et al.
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13 91 2013).

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18 93 Previous studies investigating the relationship between components of biodiversity
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20 94 and psychological well-being have focused on measuring absolute diversity (how
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22 95 much diversity is estimated actually to be present; Fuller et al. 2007, Luck et al. 2011),
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24 96 and/or the diversity that people perceive to be present (Dallimer et al. 2012, Shwartz
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26 97 et al. 2014, Belaire et al. 2015). However, these may not reflect the biodiversity that
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28 98 people actually experience. In particular, daily activity levels of people and other
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30 99 organisms often differ, so understanding the well-being effects of the diversity that
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32 100 people actually experience requires consideration of lower than actual values.
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38 102 Here we address two key questions. First, what components of nature are linked to
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40 103 positive mental health outcomes? To answer this, we explore the relationships
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42 104 between three established self-reported measures of mental health for depression,
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44 105 anxiety and stress, and five metrics of neighborhood nature (vegetation cover,
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46 106 estimated actual abundance and richness of birds, and the abundance and richness of
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48 107 birds that people are likely to experience). Our second question is whether there is a
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50 108 threshold in the mental health response. To answer this, we use dose-response
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52 109 modelling to estimate the point at which neighborhood vegetation cover (a tangible
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54 110 component of nature that relevant stakeholders can manage) influences the prevalence
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3 111 of depression, anxiety and stress, and the reduction in prevalence that could be
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5 112 achieved through enhanced exposure across the urban population.
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10 114 **Assessment of mental health and participants**

11 115 We delivered an urban lifestyle questionnaire online (see Shanahan et al. 2016 for
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13 116 details) through a market research company (Shape the Future Ltd) to 1,023 adults
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15 117 enrolled in their survey database. All participants lived within the urban limits of the
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17 118 ‘Cranfield triangle’, a region in southern England, U.K., comprising the three adjacent
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19 119 towns of Milton Keynes, Luton, and Bedford. Together they comprise an urbanized
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21 120 area of c.157 km² and an urban population of c.524,000 (2011 Census, UK). The
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23 121 triangle represents great variation in human population density (including examples of
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25 122 low and high density living), urban history and urban form. The survey was delivered
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27 123 in May 2014, a period of reasonably mild weather when respondents were most likely
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29 124 to engage with nature around their home, and so the benefits of nature may be more
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31 125 pronounced. Participants were self-selecting and were compensated with either a
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33 126 nominal fee or a prize draw entry (see supplemental appendix S1 for ethical
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35 127 clearance). A subset of 263 respondents for whom there was both vegetation and bird
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37 128 survey data, was then used in the analysis (see the metrics of neighborhood nature
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39 129 section below).
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47 131 Survey respondents were asked to complete the short version of the Depression,
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49 132 Anxiety and Stress Scale (DASS 21; Lovibond and Lovibond 1995). On a four-point
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51 133 scale respondents rated the extent to which each of 21 statements applied to them over
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53 134 the previous week (seven statements for each of depression, anxiety and stress; table
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55 135 S2a). To characterize the degree of severity for each mental disorder relative to the
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3 136 wider population, these scores were summed for each disorder before banding as
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5 137 normal, mild, moderate, severe or extremely severe (table S2b; Lovibond and
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7 138 Lovibond 1995). If a respondent did not score a statement, then the relevant disorder
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9 139 for that respondent was discarded from the analysis (remaining respondents;
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11 140 depression = 248, anxiety = 259, stress = 240).
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16 142 The survey collected socio-demographic and personal circumstance data that could
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18 143 potentially influence mental health, including age, gender, the primary language
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20 144 spoken at home, personal annual income, the number of days exercised for 30 minutes
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22 145 or more during the survey week (an indicator of physical activity), self-assessment of
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24 146 health and highest formal qualification. As a potential confound of recent nature
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26 147 exposure, we asked respondents relatively how much time they spent out of doors in
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28 148 the previous week (supplemental table S1 shows how these variables were used for
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30 149 analysis). Respondents were requested to provide a full UK postcode so that their
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32 150 neighborhood could be characterized (one UK postcode covers approximately 20
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34 151 households). Based on the postcode the English Index of Multiple Deprivation (IMD)
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36 152 was used to assess the level of socio-economic disadvantage (Sharegeo.ac.uk, data
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38 153 sourced from Data.gov.uk). Finally, using the UK gridded population based on the
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40 154 Census 2011 and Land cover map 2007 (Reis et al. 2016) we calculated neighborhood
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42 155 population density (see supplemental appendix S2 for full description of these two
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44 156 variables).
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51 158 **Metrics of neighborhood nature**

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54 159 We measured five key components of nature that people were exposed to around the
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56 160 home. We first measured neighborhood vegetation cover as vegetation ≥ 0.7 m in
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3 161 height, within a 250 m buffer around the centroid of each respondent's postcode,
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5 162 approximately reflecting the viewscape from, and the area immediately adjacent to,
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7 163 people's homes. Vegetation cover maps were derived from airborne hyperspectral and
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9 164 LiDAR (Light Detection and Ranging) data; full details of spatial product
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11 165 development are provided in the supplemental appendix S3.
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16 167 We conducted extensive bird surveys within the towns to generate a further four
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18 168 metrics of neighborhood nature. We estimated actual bird abundance and species
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20 169 richness as that recorded during early morning surveys when birds are most active and
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22 170 so most likely to be recorded (supplemental appendix S4). We also estimated the bird
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24 171 abundance and species richness that people were likely to experience, as those birds
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26 172 that were recorded during afternoon surveys when most people are also active
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28 173 (supplemental appendix S4). These were derived from point count surveys, using
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30 174 distance sampling, at up to four locations within 116 tiles, each of 500×500 m squares
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32 175 that were selected randomly across the three towns (full details are provided in
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34 176 supplemental appendix S4). We estimated neighborhood bird abundances and
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36 177 richness for those respondents whose 250 m neighborhood buffer overlapped with at
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38 178 least one bird survey location within a survey tile (respondents = 263; tiles = 84; see
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40 179 supplemental table S3 for socio-demographics of subset; supplemental figure S1
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42 180 illustrates an example of overlap). This subset of respondents was used in subsequent
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44 181 analyses.
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52 183 The neighborhood vegetation cover varied nine-fold across the 263 respondents (table
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54 184 S4). Pearson's rank sum tests of the five metrics of neighborhood nature showed that
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56 185 actual and afternoon species richness were highly correlated (Pearson's $r = 0.72$, p
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3 186 <0.0001), while the remaining nature variables were either weakly or not correlated (r
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5 187 <0.28; see supplemental table S5 for correlation matrix between nature variables).

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10 189 **Relationships between mental health and neighborhood nature**

11 190 We used ordinal regression to explore relationships between the five metrics of
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13 191 neighborhood nature and each mental health disorder in turn. We incorporated age,
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15 192 gender, language, income, physical activity, self-assessment of health, level of
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17 193 education, relative time out of doors in the previous week, neighborhood population
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19 194 density and the IMD as covariates. We standardized the five nature metrics and
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21 195 neighborhood population density such that each had mean zero and standard deviation
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23 196 one. Because multicollinearity of >0.7 can severely distort model estimation (e.g.
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25 197 Dormann et al. 2013, Cade 2015), we built two models for each mental state,
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27 198 including either actual or afternoon species richness in each along with the other three
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29 199 nature metrics and covariates. We used the ‘MuMIn’ package (Bartoń 2015) to
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31 200 produce all subsets of models based on the global model and rank them based on the
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33 201 Akaike Information Criterion (AICc). Over-dispersion in models is problematic in
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35 202 AICc analysis and may be due to not accounting for important covariates or
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37 203 multicollinearity, which can result in selection of overly complex models that can lead
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39 204 to poor inference. Following Burnham and Anderson (2002, p.131) and Richards
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41 205 (2008) we reduced the retention of overly complex models by excluding from the set
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43 206 of candidate models all models that are more complicated versions of any model with
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45 207 a lower AICc value (i.e., nesting of models). To be 95% sure that the most
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47 208 parsimonious models were maintained within the best supported model set, we then
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49 209 retained all models where $\Delta\text{AICc} < 6$ (Richards 2005, 2008). We then calculated
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3 210 averaged parameter estimates and standard errors using model-averaging among the
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5 211 retained models (Burnham and Anderson 2002).

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9 213 People living in neighborhoods with higher levels of vegetation cover and afternoon
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11 214 bird abundances had reduced severity of depression, anxiety and stress (table 1; figure
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13 215 1). In contrast, there was no relationship with the estimated actual neighborhood bird
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15 216 abundance and species richness, or afternoon species richness (table 1). Respondents
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17 217 who spent less time out of doors than usual in the last week had worse depression and
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19 218 anxiety (table 1). Respondents over the age of 45 years were less likely to suffer from
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21 219 depression than younger respondents, while those between 46 and 60 years were less
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23 220 likely to suffer from anxiety (table 1). Mental health was positively correlated with
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25 221 self-reported physical health (table 1; inherent bias within self-reported surveys is
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27 222 here, at least in part, mitigated through large sample sizes and a robust ordinal
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29 223 regression analytical approach).

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35 225 Here we have shown that metrics of nature that were most visible during the day and
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37 226 so most likely to be experienced by people, namely vegetation cover and afternoon
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39 227 bird abundances, were positively associated with a lower population prevalence of
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41 228 depression, anxiety and stress. This may have arisen for a range of non-mutually
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43 229 exclusive reasons. First, experiences of visible nature may act to improve people's
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45 230 mental health, as predicted from previous empirical studies of interactions between
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47 231 nature and well-being (see Introduction for references). Second, people with no or
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49 232 low mental health dis-orders may be self-selected by electing to move into
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51 233 neighborhoods that are greener. Third, they may provide resources for birds, thereby
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53 234 increasing opportunities for closer interactions throughout the day. Thus it is unclear
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3 235 whether a lower population prevalence of poor mental health is shaped by the natural
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5 236 environment people live in, or whether people move to a neighborhood that reflects
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7 237 that trait, or whether it is some combination of these factors. However, we found no
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9 238 relationship with the metrics estimating actual bird community composition or actual
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11 239 or afternoon species richness; nor were there relationships between mental health and
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13 240 covariates such as the IMD, education or population density, although this is not
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15 241 entirely unsurprising given the complex nature of mental health disorders and that
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17 242 previous studies have recorded wide variation in these relationships across different
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19 243 human populations (e.g. Das et al. 2007). The difference in the associations of actual
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21 244 and visible bird abundance with mental health is indicative of an effect of visible
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23 245 nature on mental health. Notwithstanding, future research needs to focus on further
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25 246 unpicking causal pathways, such as through studies of brain activity and function
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27 247 during exposure to nature (e.g. Bratman et al. 2012, 2015).
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34 249 The shape of the relationships between vegetation cover and the increasing severity of
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36 250 each mental health disorder suggests that the greatest benefits were gained by those
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38 251 respondents with mild or moderate mental health disorders (figure 1). This may be
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40 252 because the severity of depression often determines behaviors, and thus the degree to
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42 253 which people engage with nature. So people suffering from severe mental health
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44 254 disorders may be less likely to venture out of doors, and the mechanisms behind their
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46 255 disorders may be different, thereby reducing the positive influence of nature.
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49 256 Respondents who spent relatively less time out of doors in the survey week were
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51 257 more likely to report worse depression and anxiety. Intriguingly this suggests that the
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53 258 relative nature experienced is a significant contributing factor.
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3 260 We found no relationship between mental health and either measure of bird richness
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5 261 or that of actual abundance. Given that most people cannot distinguish between
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7 262 species (Dallimer et al. 2012, Shwartz et al. 2014) benefits may be provided through
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9 263 directly experiencing abundance, with richness contributing when people can see
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11 264 multiple species within a relatively small timeframe, such as around a feeder (Cox &
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13 265 Gaston 2015). Although the positive benefits for mental health of interacting with
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15 266 birds is compelling, in this study it was not possible to determine the actual
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17 267 abundances of birds that respondents interacted with and thus there may be more than
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19 268 one explanation for the positive associations between afternoon bird abundances and
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21 269 improved mental health. First, as seems likely, the abundances recorded by ecologists
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23 270 in the afternoon may be a good representation of the birds that most people
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25 271 experience and gain benefits from. Second, these abundances may be a proxy for
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27 272 another biological component.
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34 **Dose-response relationships between neighborhood vegetation cover and mental** 35 36 **health**

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38 276 We next calculated the dose-response of each mental health disorder within the
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40 277 survey population that could be attributed to levels of neighborhood vegetation cover.
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42 278 We created a further three binary response variables, those with normal mental health
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44 279 for each of depression, anxiety and stress, and those suffering with mild or worse
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46 280 cases (Lovibond and Lovibond 1995). We used logistic regression for each binary
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48 281 response variable in turn to estimate the relative odds of occurrence in an individual
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50 282 given specific risk factors that were statistically significant in the previous analysis.
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52 283 Each covariate (i.e., risk factor) was transformed into a binary factor conveying 'high'
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54 284 versus 'low' risk (see supplemental table S6). For each mental health disorder we ran
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3 285 multiple logistic regression models. The first model contained the risk factors
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5 286 described above with the binary factor vegetation cover set at 10%, below which the
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7 287 risk of poor mental health was 'high'. The model was then repeated applying an
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9 288 incrementally increased break point in vegetation cover (i.e., <15%; <20%; <25%;
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11 289 30%; <35%). We identified the point at which the health gains were first recorded as
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13 290 better than the null model on a plot of dose versus the odds ratio for use in the
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15 291 analysis described below (i.e., the confidence interval did not overlap with an odds
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17 292 ratio of one).
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23 294 For each mental health disorder we calculated the population average attributable
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25 295 fraction to estimate the proportion of cases in the population attributable to each of
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27 296 the predictor variables (or risk factors; e.g., Rueckinger et al. 2009). Each risk factor
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29 297 was removed sequentially from the population by classifying every individual as low
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31 298 risk. The probability of each person experiencing mild or worse depression, anxiety or
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33 299 stress was then calculated, where the sum of all probabilities across the population
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35 300 was the adjusted number of disease cases expected if the risk factor were not present.
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37 301 The attributable fraction was calculated by subtracting this adjusted number of cases
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39 302 from the observed number of cases. The risk factors were removed in every possible
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41 303 order, and an average attributable fraction from all analyses was obtained (table 2).
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47 305 After accounting for covariates, the odds of having mild or worse depression were
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49 306 significantly lower when neighborhood vegetation cover reached a threshold of 20%,
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51 307 with gains in the odds ratio of 0.35 by 35% vegetation cover (figure 2a). There was a
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53 308 significantly lower chance of having anxiety and stress after 30% and 20% vegetation
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55 309 cover respectively, although there was greater variability in the dose-response curve
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3 310 (figure 2b and 2c). The power of the tests for all three mental health disorders was
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5 311 reduced at higher levels of vegetation cover (indicated by wider 95% confidence
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7 312 intervals) because the proportion of respondents reporting poor mental health declined
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10 313 at these levels; increasing the number of respondents may reduce the variability in the
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12 314 dose-response curves.

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16 316 This threshold analysis has important implications for setting future research
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18 317 directions towards operationalizing nature as a tool for improving health and well-
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20 318 being for populations. While there is unlikely to be a 'one size fits all' policy for
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22 319 optimizing nature in cities, establishing minimum levels of vegetation cover in
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24 320 neighborhoods is a practical approach that could be incorporated into city design.

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29 322 The results suggest that if all respondents lived in neighborhoods with vegetation
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31 323 cover of >20% then the total number showing symptoms of depression would be
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33 324 reduced by up to 11%. The number of cases of anxiety and stress could be reduced by
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35 325 up to 25% and 17% if vegetation cover were >30% and >20%, respectively. Within
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37 326 the survey population 38%, 76% and 38% of respondents were considered at risk of
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39 327 showing symptoms of depression, anxiety and stress respectively, because
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41 328 neighborhood vegetation cover levels were not met. In 2007 it was estimated that
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43 329 depression cost the English economy £7.5 billion and anxiety cost £8.9 billion in
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45 330 health costs and lost workdays (McCrone et al. 2008). Although the causes of poor
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47 331 mental health are diverse, a simplistic calculation would be that if minimal levels of
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49 332 neighborhood vegetation cover were met, it has the potential to contribute towards an
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51 333 annual saving of up to £0.5 and £2.6 billion per year for depression and anxiety alone.
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53 334 Doubtless the financial implications are marked. Consequently, manipulation of
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3 335 neighborhood vegetation and bird populations to ‘optimal’ levels can and should be
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5 336 encouraged to be undertaken by both private and public stakeholders. There are
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7 337 multiple approaches available such as through the innovative addition of green
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9 338 infrastructure like tree planting, the addition of green walls and roofs (Tzoulas et al.
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11 339 2007), or the provision of supplementary food and nesting locations to increase local
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13 340 bird abundances (Fuller et al. 2008) and bring birds into closer contact with people.
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18 342 Research is starting to tease apart the mechanistic pathways behind how different
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20 343 components of nature benefit mental health (e.g., Bratman et al. 2015, Shanahan et al.
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22 344 2015a). Although this study does not demonstrate causation per se, the positive
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24 345 relationships between two metrics of neighborhood nature and better mental health
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26 346 are consistent with a mechanistic effect. Indeed, the dose-response relationship for
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28 347 depression, and to a lesser extent anxiety and stress, is considered to provide some
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30 348 evidence of causality according to Hill’s criteria (Hill 1965). These benefits are likely
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32 349 to be provided via two pathways, first by increasing the attractiveness and appeal of
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34 350 green space such that people are more likely to spend time out of doors and thus
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36 351 increase the likelihood that they will engage in physical or social activities, and
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38 352 second increasing the visual complexity of the landscape enhancing its effect on
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40 353 mental restoration and well-being (Shanahan et al. 2015b). However, at the same time
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42 354 it is important to acknowledge that living close to too much, or inappropriate, nature
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44 355 can also provide a range of dis-services such as destruction of property from
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46 356 vegetation and breeding birds (e.g. Rock 2005), or increased levels of vegetation
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48 357 leading to feelings of decreased safety in some neighborhoods (e.g. Kuo et al. 1998).
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50 358 Future research into ‘best’ doses of nature would benefit from exploring the trade-offs
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52 359 between the benefits and dis-services.
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361 **Conclusion**

362 Although the causes and drivers of poor mental health are diverse (Kinderman et al.
363 2015) this study suggests that even low levels of key components of neighborhood
364 nature can be associated with better mental health, providing promise for preventative
365 health approaches. This study shows that quantifiable reductions in the population
366 prevalence of poor mental health can be achieved if minimal thresholds of vegetation
367 cover are met. This has important implications for policy to set minimum levels of
368 neighborhood nature and paves the way to test for health gains that arise from specific
369 interventions. Obviously, optimized levels of nature are not a silver bullet for the
370 prevention or treatment of mental health problems, but it is an approach that can and
371 should be applied in conjunction with existing frameworks such as medical and social
372 services, reducing crime and increasing community driven action. Indeed, optimizing
373 the key components of nearby nature have been shown to change behavior towards
374 increased social cohesion (e.g. Weinstein et al. 2015) and green exercise (e.g.
375 Mitchell and Popham 2008).

376

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3 385 available on request from the corresponding author, and will be made available from
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5 386 mid 2017 at the NERC Environmental Data Information Centre.
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529 **Table 1:** Nested model averaging of ordinal regression showing negative
 530 relationships between two visible components of nature around the home and three
 531 mental health disorders, whilst adjusting for socio-demographic factors. For the
 532 categorical variables (listed in italics), we show the model-averaged coefficients (with
 533 standard errors in brackets) of variables relative to a comparative base factor level
 534 (e.g., age ≤ 30 years, so a positive coefficient suggests that those >30 years have worse
 535 mental health; other base factors are: Gender, females; Language, English is the
 536 primary language spoken at home; Relative time outdoors, Less time; Self-assessment
 537 of health, very poor; Education, 16+ education).

Variables	Depression	Anxiety	Stress
Vegetation cover	-0.41 (0.15)**	-0.34 (0.16)*	-0.30 (0.15)*
Actual abundance	-	0.26 (0.16)	0.25 (0.16)
Actual richness ⁺	-	-	-
Afternoon abundance	-0.43 (0.15)**	-0.54 (0.18)**	-0.35 (0.18)*
Afternoon richness ⁺	-	-	-
Gender (male)	-	0.49 (0.32)	-
Language	0.57 (0.36)	-	-
Income	0.13 (0.08)	-	-
Physical activity	-	-	-
IMD	-	-	-
Population density	-	-	-
<i>Relative time outdoors</i>			
About the same	-0.74 (0.33)*	-0.95 (0.36)*	0.55 (0.36)
More time	-0.84 (0.38)*	-1.29 (0.42)**	-0.88 (0.48)
<i>Age</i>			
Age (31 to 45 yrs)	-0.11 (0.33)	-0.02 (0.35)	0.59 (0.35)
Age (46 to 60 yrs)	-1.13 (0.39)**	-1.23 (0.44)**	-0.78 (0.46)
Age (>60 yrs)	-1.90 (0.82)*	-0.93 (0.65)	-1.70 (1.07)
<i>Self-assessment of Health</i>			
Poor	-1.81 (1.02)	-3.75 (1.39)**	-
Average	-2.28 (0.94)**	-3.92 (1.32)***	-
Good	-3.49 (0.95)***	-4.57 (1.32)***	-
Very good	-3.30 (0.96)***	-4.73 (1.35)***	-
<i>Level of education</i>			
Education (18+)	-	-	-
Education (Undergraduate)	-	-	-
Education (Postgraduate)	-	-	-

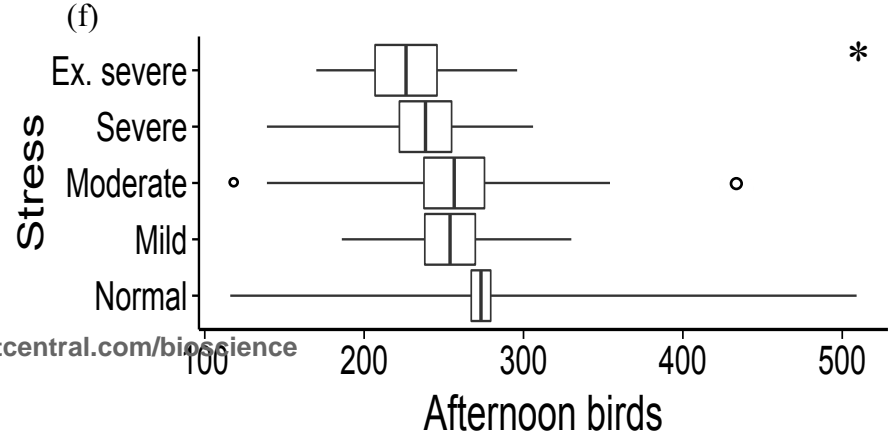
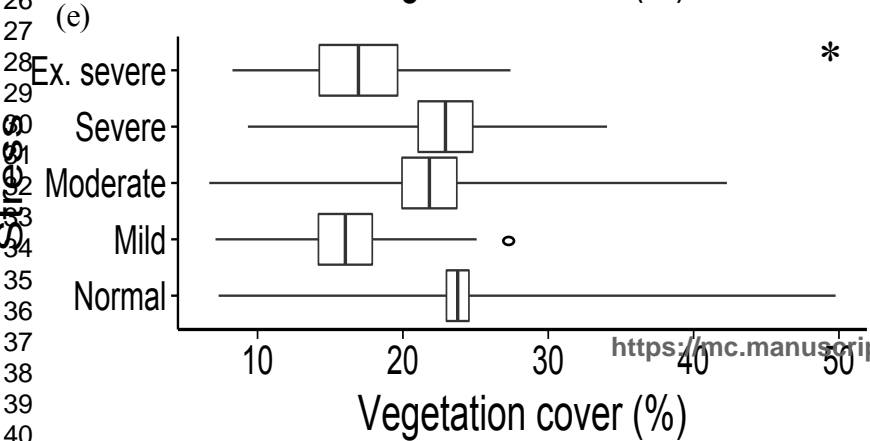
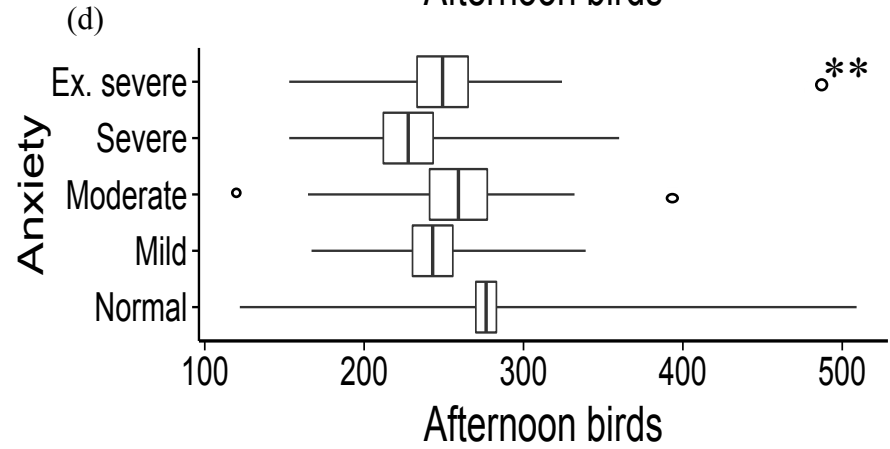
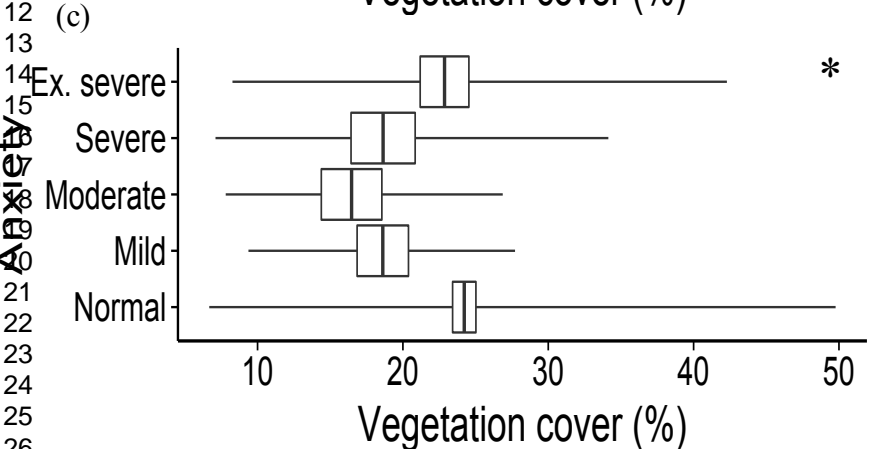
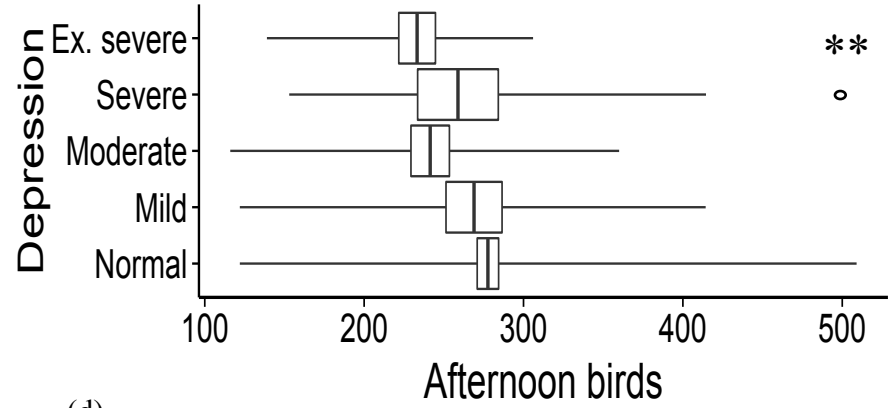
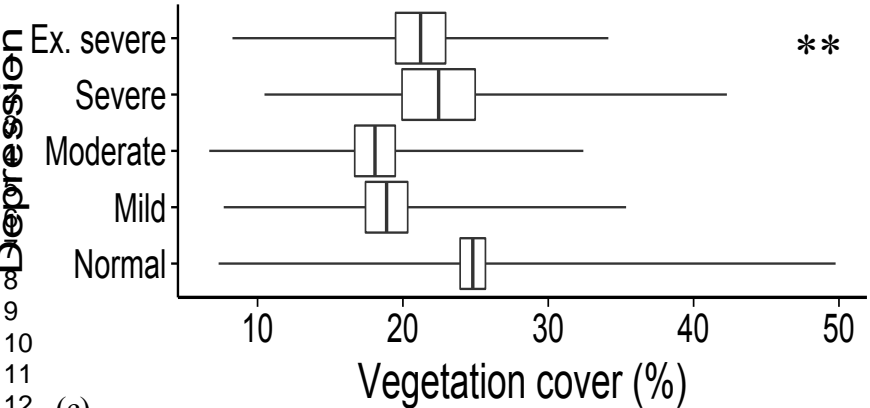
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3 538 Significant variables and factor levels relative to base level are shown as: *P < 0.05;
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5 539 **P < 0.01; ***P < 0.001. [†] For each mental health disorder we built two identical
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7 540 models, testing each measure of richness separately (see methods) – variable was not
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10 541 retained in the top nested models where delta <6.
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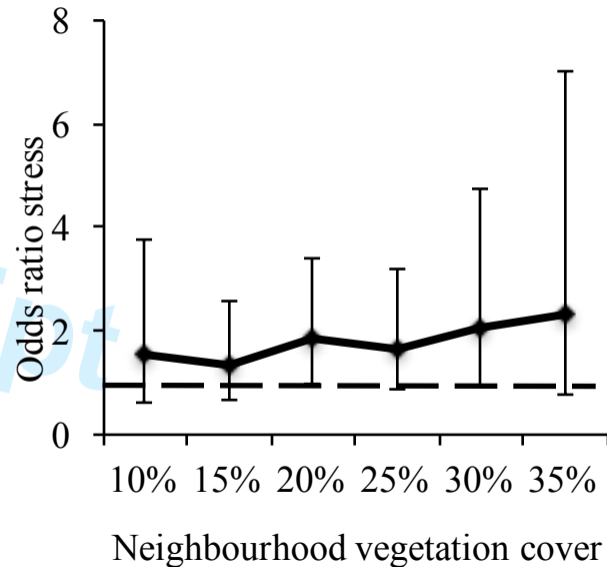
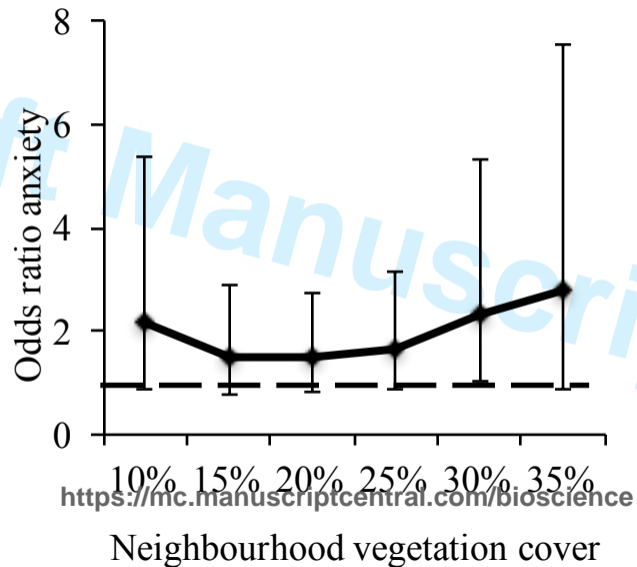
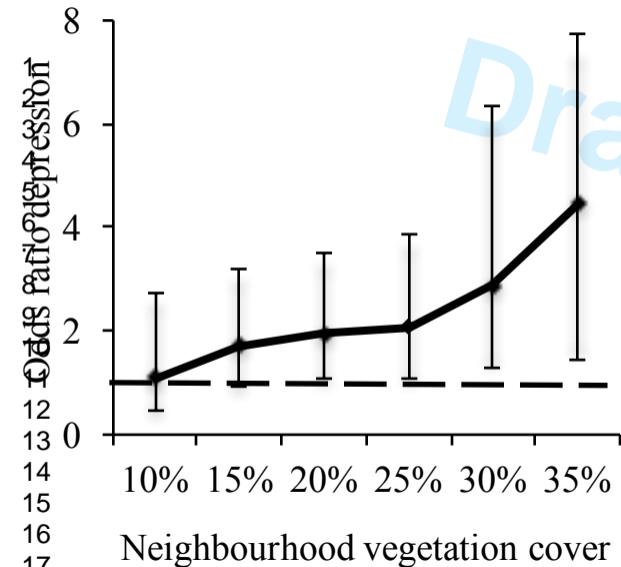
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542 **Table 2:** Dose-response modelling shows the proportion of mental health cases in the study population attributable to various risk factors
 543 (average population attributable fraction; AAF), we show a positive association between a reduced population prevalence of depression, anxiety
 544 and stress and minimal thresholds of neighborhood vegetation cover* (depression >20% cover, anxiety >30% cover, stress >20% cover). An
 545 odds ratio above 1 indicates the mental health disorder is more likely to be present where the risk factor is present.

	Risk factor	Depression		Anxiety		Stress	
		Odds ratio (95% CI)	AAF	Odds ratio (95% CI)	AAF	Odds ratio (95% CI)	AAF
Age	Higher risk <46 years	3.28 (1.48:7.78)	0.37	2.11 (0.96:4.66)	0.22	NA	NA
Self-assessment of health	Higher risk <average health	6.0 (2.02:17.8)	0.07	3.65 (1.31:9.98)	0.49	NA	NA
Relative time outdoors	Higher risk <less time outdoors	3.30 (1.64:6.62)	0.12	2.50 (1.25:4.98)	0.09	NA	NA
Vegetation cover	Higher risk <% veg. cover*	2.00 (1.11:3.61)	0.11	2.29 (1.01:5.20)	0.25	1.76 (1.01:3.83)	0.17
Afternoon bird abundance	Higher risk <266 birds	2.03 (1.16:3.52)	0.15	3.05 (1.70:5.50)	0.24	1.70 (0.93:3.44)	0.17

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Supplemental Material

Doses of Neighborhood Nature: Benefits for Mental Health of Living with Nature.

DTC Cox, DF Shanahan, HL Hudson, KE Plummer, GM Siriwardena, RA Fuller, K

Anderson, S Hancock & KJ Gaston.

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Appendix S1. Ethical clearance

This research was conducted with approval from the Bioscience ethics committee of the University of Exeter (project number 2013/319). Participants provided written consent at the beginning of the online survey by checking a box stating their agreement to participate.

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3 **Appendix S2. Calculation of Index of Multiple Deprivation and neighborhood**
4 **population density.**
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7 *Index of multiple deprivation:* We used the Indices of Multiple Deprivation (IMD,
8 Sharegeo.ac.uk, data sourced from Data.gov.uk) produced by communities and local
9 government to derive a socio-economic deprivation score for each tile. The IMD contained
10 separate indices for separate domains of deprivation (e.g. ward level income, employment,
11 health deprivation and disability), which were simply averaged. This IMD is provided at the
12 postcode scale.
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22 *Neighborhood population density:* We used the UK gridded population based on the Census
23 2011 and Land cover map 2007 (Reis et al. 2016). This dataset contains a gridded population
24 density with a spatial resolution of 1 km x 1 km. For each 250 m buffer around the centroid
25 of a respondent's postcode we scaled the estimated population relative to the area of the
26 gridded population square that the buffer covered. Where the buffer covered more than one
27 tile we weighted our estimate by the proportion of each tile that was covered.
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3 **Appendix S3. Characterization of the urban form for each tile.** The urban form for each
4 tile was characterised using airborne hyperspectral (Eagle imaging spectrometer; 12 bit,
5 pushbroom, hyperspectral sensor with a 1000 pixel swath width, covering the visible and near
6 infra-red spectrum 400 - 970nm) and LiDAR (Light Detection and Radar) (Leica ALS50-II)
7 data collected by the Natural Environment Research Council (NERC) Airborne Research and
8 Survey Facility (ARSF) aircraft in July and September 2012. The normalized difference
9 vegetation index (NDVI; Tucker 1979) was calculated from the hyperspectral data using a
10 red band centered on 570nm and a near infra-red band centered on 860 nm with a spatial
11 resolution of 2m. Histograms of NDVI were examined and a threshold of 0.2 identified as
12 being suitable to separate vegetated ($NDVI \geq 0.2$) from non-vegetated ($NDVI < 0.2$) pixels
13 (Liang 2004). The LiDAR data were used in discrete return mode, with up to four returns per
14 laser pulse. The laser point density was between one point per 25cm^2 and one point per 2m^2 ,
15 depending on flight line overlap. The lastools 'lasground' function (Isenburg 2011) was used
16 to find ground returns within the LiDAR point cloud. Pixels (2m resolution) with an NDVI
17 greater than 0.2 and a mean height of first return more than 0.7 m above the ground were
18 marked as trees (figure S1). Heights from discrete return LiDAR are well-known to produce
19 biased results over vegetation (Hancock et al. 2011) and so this 0.7 m threshold may have
20 represented a more variable vegetation threshold height, and since that bias is most usually an
21 underestimation, it could correspond to taller vegetation (up to 1.7 m tall). We then measured
22 neighborhood tree cover as vegetation ≥ 0.7 m in height, within a 250 m buffer around the
23 centroid of each respondent's postcode. We estimated that the average area of the
24 respondents' postcodes was $12,436\text{ m}^2$, (14,257 SD), and so fell within the neighborhood
25 buffer.
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3 **Appendix S4. Estimation of neighborhood actual bird abundance and species richness,**
4 **and the abundance and richness of birds that people are likely to experience.** We divided
5 the landscape within the urban limits of the three towns into 500×500 m square ‘tiles’
6 (250,000 m²), where tiles within the urban limit were defined as those within the
7 administrative boundary that had greater than 25% urban built form, as assessed by eye
8 (Gaston et al. 2005). As a measure of urban form within each tile we calculated the tree cover
9 (Appendix S3), as well as the number of building polygons shown in the Ordnance Survey
10 MasterMap (2013). We then selected a subset of 116 tiles using random sampling stratified to
11 provide consistent variation in urban form between 0% and 50% tree cover and between 0
12 and 20% built cover, reflecting the range of values found in the study towns. Within each tile,
13 up to four point locations were identified (mean per tile, 3.89 ± 0.37 SD), in order to
14 represent the diversity of urban forms present as fully as possible, subject to access
15 restrictions. Survey points were selected to be ≥200 m apart and ≥100 m from tile edges, such
16 that surveys from each point sampled different birds, with fewer than four points being
17 chosen if sufficiently separated points could not be accessed.
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39 To measure bird abundance, we used point counts and a distance sampling procedure to
40 account for differences in detectability among species (Buckland et al. 2001). All point
41 counts were conducted by one of two trained researchers. To estimate actual bird abundance,
42 two early-morning surveys (06:00 – 10:00 hours) were conducted at each point in all tiles
43 during the breeding season, one in May and one in June; these were timed to maximize the
44 detectability of the component species of the local breeding bird community. The abundance
45 that people are likely to experience was estimated using two later-day surveys (10:00 – 18:00
46 hours) that were conducted in each tile from May – July using the same protocols as the
47 early-morning surveys. Point counts were conducted for ten-minute periods, divided into
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3 two-minute intervals. Within each two-minute interval, the number of birds and the radial
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5 distance from the observer at which they were seen was recorded in bands of 0-20 m, 20-40
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7 m, 40-60 m and 60-100 m. Birds were recorded independently in each two-minute period.
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9 Individuals that moved during a two-minute period were recorded in the distance band in
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11 which they were first detected.
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16 For each individual ten-minute point count, we selected the maximum count of each species
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18 per distance band from the multiple two-minute intervals. For each band we then selected the
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20 maximum total count across visits. We used these data and the 'unmarked' package (Fiske
21
22 and Chandler 2011) to calculate bird abundance corrected for detection probability, whilst
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24 adjusting for percentage tree cover >0.7 m and percentage built cover within the survey
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26 radius. We calculated a pooled detection function for species with similar morphology and
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28 behavior, assuming that these species had similar detection characteristics (table S7). This is
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30 because a number of species had small sample sizes (<40 records), which precluded
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32 appropriate distance analysis on these individual species. Species with small sample sizes that
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34 were morphologically and/or behaviorally distinct were excluded from analysis (Northern
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36 lapwing *Vanellus vanellus*; pheasant *Phasianus colchicus*; European cuckoo *Cuculus*
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38 *canorus*; European kingfisher *Alcedo atthis*; Grey wagtail *Motacilla cinerea*). Models failed
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40 to converge for a further five groups of species and the original abundances were used (table
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42 S7). Because detection of species might vary with time of day (Alldredge et al. 2007) we
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44 calculated detection probabilities for each species, or group of species, for both early morning
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46 and afternoon surveys. We calculated an adjusted measure of abundance in each survey tile
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48 by dividing raw abundance counts for each species by its detection probability, before
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50 summing adjusted counts across species (or unadjusted counts for those species or groups of
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3 species where models failed to converge) within each survey point to get the total abundance
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5 by survey tile.
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10 To estimate actual species richness within each survey tile we calculated the total number of
11 species seen across all early morning surveys. We repeated this for afternoon surveys to
12 obtain an estimate of the species richness that people are likely to experience.
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Table S1. Socio-demographic variables used in the analysis. Variables used to examine the relationship between three negative mental states (depression, anxiety and stress) and a participant's exposure to five metrics of nature intensity.

Name	Variable	Description	Supporting reference for inclusion
Age (ordinal)	Age	Respondents selected from 11 brackets: 18-20 years, then increasing increments of five years until >60 years. Responses were then banded as: 18-30 years; 31-45 years; 46-60 years; >60 years	(Astell-Burt et al. 2014a, Mroczek and Kolarz 1998, Wu et al. 2015)
Gender (categorical)	Gender	Female or Male	(Rosenfield and Mouzon 2013, Ross and Mirowsky 2006)
Language (categorical)	Primary language spoken at home	Respondents speak a language other than English at home (No or Yes)	(Bratter and Eschbach 2005, Wu et al. 2003)
Income (categorical)	Personal annual income	Respondents selected from eight brackets: No income; £1-£199 a week (£1-£10,399 per year); £200-£299 a week (£10,400-£10,599 per year); £300-£399 (£15,600-£20,799 per year); £400-£599 a week (£20,800-£31,199 per year); £600-£799 a week (£32,200-£41,599 per year); £800-£9,999 a week (£41,600-£51,999 per year); >£1,000 a week (>£52,000)	(Astell-Burt et al. 2014b, de Vries et al. 2003, Weich et al. 2001)
Physical activity (numeric)	Physical activity	Self-reported number of days in previous week that the respondent exercised for more than 30 minutes	(Barton and Pretty 2010, Cohen-Cline et al. 2015, Deslandes et al. 2009, Mitchell 2013, Richardson et al. 2013)
Physical health (ordinal)	Self-assessment of health	Respondents selected from: Very poor; poor; average; good; very good	(Maas et al. 2006, van Dillen et al. 2012)
Education (categorical)	Highest formal education	Highest qualification (selected from four categories equivalent to: General Certificate of Secondary Education (GCSE); A-levels; Bachelor's degree; Postgraduate degree)	(Fryers et al. 2003, Miech and Shanahan 2000)
Recent nature experience (factor)	Relative time spent in nature in previous week	Respondents selected from: less time; about the same time; more time.	
IMD (numeric)	Index of Multiple Deprivation	Indices of Multiple Deprivation weighted for the 250m buffer of the centroid of the respondents postcode (see appendix S2).	IMD, Sharegeo.ac.uk , data sourced from Data.gov.uk
Population density (numeric)	Population density	UK gridded population based on the census 2011 and Land cover map 2007, weighted for the 250m buffer of the centroid of the respondents postcode (see appendix S2).	(Reis et al. 2016)

Table S2. Depression, Anxiety and Stress Scale 21 (DASS 21) included in the urban lifestyle questionnaire (taken from (Lovibond and Lovibond 1995) and reproduced here for ease of reference). Seven statements rated each mental health state. a) Answers to each statement were given on a four-point scale from: did not apply to me at all; applied to me to some degree, or some of the time; applied to me to a considerable degree, or a good part of the time; applied to me very much, or most of the time. b) The severity of each mental health state was then rated by summing the relevant scores.

a) Statement	Mental state
I found it hard to wind down	Stress
I was aware of dryness of my mouth	Anxiety
I couldn't seem to experience any positive feeling at all	Depression
I experienced breathing difficulty (e.g. excessively rapid breathing, breathlessness in the absence of physical exertion)	Anxiety
I found it difficult to work up the initiative to do things	Depression
I tended to over-react to situations	Stress
I experienced trembling (e.g. in the hands)	Anxiety
I felt that I was using a lot of nervous energy	Stress
I was worried about situations in which I might panic and make a fool of myself	Anxiety
I felt that I had nothing to look forward to	Depression
I found myself getting agitated	Stress
I found it difficult to relax	Stress
I felt down-hearted and blue	Depression
I was intolerant of anything that kept me from getting on with what I was doing	Stress
I felt I was close to panic	Anxiety
I was unable to become enthusiastic about anything	Depression
I felt I wasn't worth much as a person	Depression
I felt that I was rather touchy	Stress
I was aware of the action of my heart in the absence of physical exertion (e.g. sense of heart rate increase, heart missing a beat)	Anxiety
I felt scared for no good reason	Anxiety
I felt life was meaningless	Depression

b)	Depression	Anxiety	Stress
Normal	0 – 4	0 – 3	0 – 7
Mild	5- 6	4- 5	8 – 9
Moderate	7 – 10	6 – 7	10 – 12
Severe	11 – 13	8 – 9	13 – 16
Extremely severe	14 +	10 +	17 +

Table S3. Distribution of the subset of respondents for which we calculated metrics of neighborhood nature, across socio-demographic variables within the study towns (263 respondents). For comparison we also show the distribution of the Buckinghamshire and Bedfordshire counties, 2011 Census population average.

Variable	Level	Subset of survey respondents	Local population
Gender	Female	56%	51%
	Male	44%	49%
Income	No income	5%	5%
	£1-£10,399 per year	12%	12%
	£10,400-£10,599 per year	11%	19%
	£15,600-£20,799 per year	14%	11%
	£20,800-£31,199 per year	26%	25%
	£32,200-£41,599 per year	16%	18%
Age	>£41,600	15%	10%
	18-30 years	29%	21%
	31-45 years	35%	25%
	46-60 years	27%	32%
	>60 years	9%	21%
English is not the primary language spoken at home	No	85%	72%
	Yes	15%	28%
Self-assessment of health	Very poor	1.5%	0.8%
	Poor	5.7%	2.7%
	Average	27.0%	10.7%
	Good	40.3%	33.8%
	Very good	24.7%	52%
Highest level of education (or equivalent)	16+ (Secondary)	18%	28%
	18+ (A-level)	40%	12%
	Undergraduate	33%	27%
	Postgraduate	9%	8%
Physical activity (> 30 minutes exercise a week)*	0 days	29%	-
	1 day	19.4%	-
	2 days	16%	-
	3 days	14%	-
	4 days	8%	-
	5 days	9%	-
	6 days	1%	-
	7 days	4%	-
Relative time spent out of doors in previous week*	Less time	-	-
	About the same	-	-
	More time	-	-

* data is unavailable for county averages

Table S4. We show a) variation in five metrics of neighborhood nature, and b) a count of respondents for the severity for each mental health disorder.

a) Variable	Mean	Range
<i>Metrics of neighborhood nature</i>		
Vegetation cover (%)	23 (± 10)	6-50
Bird actual abundance	541 (± 100)	254-886
Bird actual species richness	22 (± 4)	14-33
Bird afternoon abundance	267 (± 79)	116-509
Bird afternoon species richness	15 (± 3)	10-23

b) Mental health	Normal	Mild	Moderate	Severe	Ex. severe
Depression	148	22	29	20	23
Anxiety	178	18	15	17	30
Stress	182	14	24	14	8

Table S5. Pearson's correlations between five metrics of neighborhood nature. For comparison we also show correlations with vegetation cover within the bird survey tiles.

Nature metric	Vegetation cover	Actual abundance	Actual richness	Afternoon abundance	Afternoon Richness
Vegetation cover	-				
Actual abundance	0.08	-			
Actual richness#	0.08	0.19	-		
Afternoon abundance	0.11	0.29	0.05	-	
Afternoon richness	0.14	0.23	0.72	0.27	-
Vegetation cover in bird survey tiles	0.57	-0.07	-0.08	0.11	-0.00

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Table S6. Binary risk factors for each covariate that was significant in the first analysis.

Variable	Conversion to binary risk factor
Age	The prevalence of mood disorders begin to decline in Australia at around 45 years (Statistics 2009). We therefore created a binary risk factor, at which above 45 years the risk of having poor mental health was zero and below was one.
Self-assessment of health	There is a higher prevalence of poor mental health in people with poor physical health (e.g. Osborn 2001). We created a binary risk factor at which the risk of having poor mental health was zero in people with average to very good health, and one in people with poor to very poor health.
Relative time spent out of doors in previous week	No information was available on time spent out of doors in the previous week and mental health. We thus considered that people had an increased risk if they spent less time out of doors in the previous week than usual.
Afternoon bird abundances	No information was available on bird abundances and mental health. We thus considered that people had an increased risk if they resided in a neighborhood with afternoon bird abundances below the median neighborhood bird abundance of this study (266 individual birds).
Neighborhood vegetation cover	We created multiple binary risk factors in increasing increments of 5%, for break points of neighborhood tree cover (10%, 15%, 20%, 25%, 30%, 35%). Where the levels of neighborhood vegetation cover were below this break point the risk of poor mental health was one, and above the risk was zero.

Table S7. Pooled detection functions for species with similar morphology and behavior.

We show whether models converged (there was no difference in convergence between morning and afternoon surveys).

Species group	Species	Model converged?
Duck	Great crested grebe (<i>Podiceps cristatus</i>); Cormorant (<i>Phalacrocorax carbo</i>); Mute swan (<i>Cygnus olor</i>); Greylag goose (<i>Anser anser</i>); Canada goose (<i>Branta canadensis</i>); Mallard (<i>Anas platyrhynchos</i>); Moorhen (<i>Gallinula chloropus</i>); Coot (<i>Fulica atra</i>)	No
Raptor	Red kite (<i>Milvus milvus</i>); Buzzard (<i>Buteo buteo</i>); Sparrowhawk (<i>Accipiter nisus</i>); Kestrel (<i>Falco tinnunculus</i>); Hobby (<i>Falco subbuteo</i>); Peregrine (<i>Falco peregrinus</i>)	No
Wader	Little egret (<i>Egretta garzetta</i>); Grey heron (<i>Ardea cinerea</i>); Oyster catcher (<i>Haematopus ostralegus</i>); Common sandpiper (<i>Actitis hypoleucos</i>); Green sandpiper (<i>Tringa ochropus</i>)	No
Gull	Black-headed gull (<i>Chroicocephalus ridibundus</i>); Herring gull (<i>Larus argentatus</i>); Lesser black-backed gull (<i>Larus fuscus</i>); Common gull (<i>Larus canus</i>)	No
Woodpecker	Green woodpecker (<i>Picus viridis</i>); Great spotted woodpecker (<i>Dendrocopos major</i>)	Yes
Flier	Swift (<i>Apus apus</i>); Barn swallow (<i>Hirundo rustica</i>); House martin (<i>Delichon urbicum</i>)	No
Thrush	Song thrush (<i>Turdus philomelos</i>); Mistle thrush (<i>Turdus viscivorus</i>)	Yes
Warbler	Garden warbler (<i>Sylvia borin</i>); Blackcap (<i>Sylvia atricapilla</i>); Whitethroat (<i>Sylvia communis</i>); Lesser whitethroat (<i>Sylvia curruca</i>); Sedge warbler (<i>Acrocephalus schoenobaenus</i>); Reed warbler (<i>Acrocephalus scirpaceus</i>); Willow warbler (<i>Phylloscopus trochilus</i>); Meadow pipit (<i>Anthus pratensis</i>); Reed bunting (<i>Emberiza schoeniclus</i>)	Yes
Tit	Goldcrest (<i>Regulus regulus</i>); Coal tit (<i>Periparus ater</i>); Marsh tit (<i>Poecile palustris</i>); Long-tailed tit (<i>Aegithalos caudatus</i>); Nuthatch (<i>Sitta europaea</i>); Tree creeper (<i>Certhia familiaris</i>); Bullfinch (<i>Pyrrhula pyrrhula</i>)	Yes
Finch	Goldfinch (<i>Carduelis carduelis</i>); Greenfinch (<i>Chloris chloris</i>); Yellowhammer (<i>Emberiza citrinella</i>); Linnet (<i>Linaria cannabina</i>)	Yes
Corvid	Magpie (<i>Pica pica</i>); Jay (<i>Garrulus glandarius</i>); Jackdaw (<i>Corvus monedula</i>); Rook (<i>Corvus frugilegus</i>); Carrion crow (<i>Corvus corone</i>)	Yes

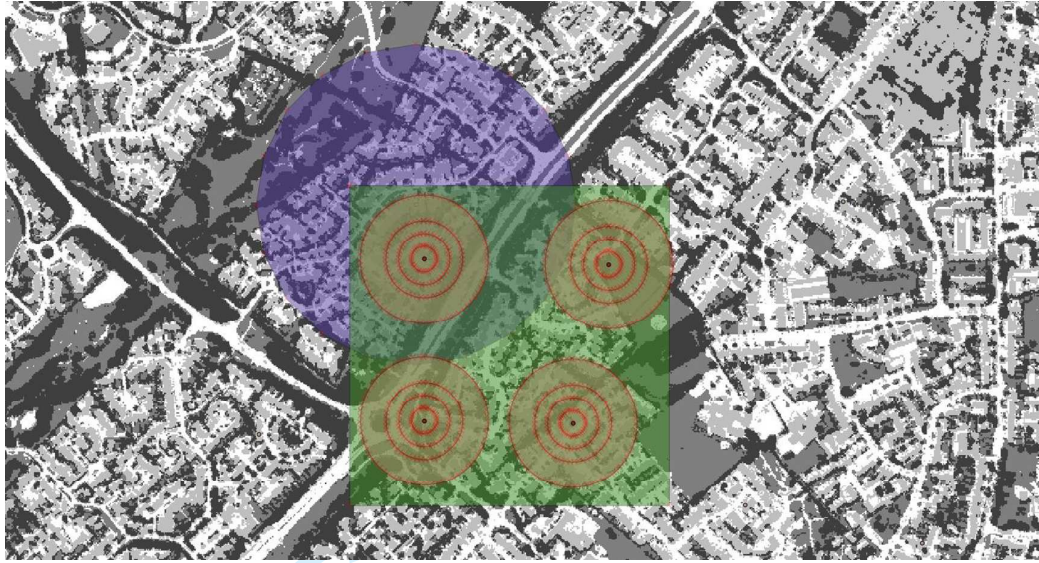


Figure S1. Remote sensing image showing how neighborhood bird abundances and richness were estimated for those respondents whose 250m-neighborhood buffer (purple) overlapped with at least one bird survey location (brown) within a survey tile (green). We show remote sensing land classifications within each urban area (dark grey, trees >0.7 m; medium grey, grass and shrubs <0.7 m; light grey, no vegetation >0.7 m; no vegetation ,0.7 m).

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Statement of Author Contributions

D.T.C.C and K.J.G conceived and designed the study. DFS, RAF and KJG designed and wrote the urban lifestyle survey. KEP and GMS managed and collated the bird survey data. KA and SH produced the remote sensing layers. DTCC carried out the analysis and wrote the paper. HLH helped write the paper. All authors edited the paper.

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