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# 1 The relationship between neighbourhood walkability and

## 2 cardiovascular risk factors in northern France

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# 12 Abstract

Background. Although walkability is known to be associated with obesity and hypertension through
increased physical activity; data on cardiovascular risk factors (especially in the Europe) are scarce.
We assessed the relationship between neighbourhood walkability and cardiometabolic factors
(including obesity, hypertension, the blood lipid profile, and serum glycated haemoglobin (HbA1c)
levels) among adults living in northern France.

Methods. Data were extracted from the ELISABET study database (2011-2013). The participants (aged between 40 and 65) resided in or around the cities of Lille and Dunkirk. For each residential address, we determined a neighbourhood walkability index (using a geographic information system) and the Walk Score<sup>®</sup>. Multilevel linear and logistic models were used to assess the relationships between neighbourhood walkability on one hand and body mass index (BMI), obesity, blood pressure, hypertension, serum HDL-C, LDL-C, triglyceride and HbA1c levels, and physical activity level
on the other.

25 Results. 3218 participants were included. After adjusting for individual and neighbourhood variables, we found that a higher neighbourhood walkability index was associated with a lower BMI (-0.23 26 27 kg.m<sup>-2</sup>; 95% confidence interval (CI) [-0.44;-0.01] for a one interquartile range (IQR) increment), a 28 lower systolic blood pressure (-1.66 mmHg; 95% CI [-2.46;-0.85] per IQR), a lower prevalence of 29 hypertension (% of increase : -7.12, 95% CI [-13.56;-0.52] per IQR), and a higher prevalence of 30 moderate or high physical activity (% of increase = 6.9; 95% CI [1.2;12.72] per IQR). The walkability 31 index was not significantly associated with other cardiovascular risk factors. Similar results were 32 observed for the Walk Score<sup>®</sup>.

Conclusion. Our results showed that residence in a more walkable neighbourhood was associated
 with a lower prevalence of vascular risk factors. Promoting neighbourhood walkability might help to
 improve the population's cardiovascular health.

36 <u>Keywords:</u> walkability; built environment; vascular risk factor; obesity; physical activity

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#### 38 <u>Highlights :</u>

- Walkability is a composite measure of how walking-friendly an environment is.
- We assessed associations between walkability and cardiovascular risk factors.
- We studied data (n=3218) from a cross-sectional health survey in northern France.
- Walkability was associated with higher physical activity, and lower blood pressure and BMI.
- Modifying the built environment might improve cardiovascular health.

#### 45 I. Introduction

46 Cardiovascular disease (CVD) is the leading cause of death worldwide and remains a 47 significant public health problem in industrialized countries(1). In France, cardiovascular disease is 48 the second-leading cause of death after cancer; in 2017, almost 65,000 deaths were due to CVD (1). 49 Worldwide, more than 70% of CVD cases can be attributed to modifiable risk factors, such as diet and lifestyle (2). The French ESTEBAN study found prevalences of 17% for obesity (3), 30.6% for 50 51 hypertension (4) and 5% for treated diabetes (5). A sedentary lifestyle and unhealthy food choices 52 are among the main causes of CVD and are often related to socioeconomic and environmental 53 variables (6).

Cardiovascular disease and low levels of physical activity are interrelated (7–9). The World Health Organization (WHO) recommends doing at least 150 minutes of moderately intense physical activity or 75 minutes of vigorous physical activity per week (10). However, nearly 40% of the French population report that they do not follow these recommendations (3). A better understanding of the determinants of physical activity (including travel-related and leisure-time physical activities) might help to better promote healthier behaviour, with physical activity performed throughout the day and at low cost.

61 There is growing interest in environmental initiatives that might prevent CVD. Recent studies have shown that certain characteristics of the built environment (including walkability) are 62 63 associated with greater levels of transport-related and leisure-time physical activity (11,12). 64 Neighbourhood walkability is a composite measure of how friendly an environment is for walking; it 65 comprises objectively measured parameter. The most frequently used are residential density, street 66 connectivity, and land use diversity (13). A growing body of research suggests that walkability is 67 related to cardiometabolic health outcomes (14); indeed, a recent meta-analysis found strong 68 evidence for longitudinal relationships between neighbourhood walkability on one hand and obesity, 69 hypertension, and type 2 diabetes on the other (15). Few examined the relationship with other 70 cardiometabolic health outcomes (e.g. cholesterol and triglyceride (TG) levels)(16-18). However,

71 most of the studies were performed in North America and those associations have rarely been assessed in a European context. Indeed, effect of walkability on health may be impacted by different 72 73 pattern of urbanism in Europe and America. Historically, the medieval European city was developed 74 radiating out from the centre. Conversely, American cites were planned with essential role given to 75 cars and with low density. Europe is characterized by the relative importance of large and medium-76 sized cities and by a marked proximity between cities(19). Cars are most frequently used for 77 transportation in US cities than in European cities(20). The European cities are characterized by 78 higher structural variability, than American and some Asian cities, and by a decreasing uniform 79 concave density gradient from the centre to the periphery, compared to a relative increase in density 80 at the periphery of American cities .(21)

81 Although the European results appears to be consistent with the North American results for 82 hypertension (22,23), they rather indicate a lack of association with type 2 diabetes and obesity 83 (24,25). This is probably due to specific features of European urban settings and dietary patterns. To 84 the best of our knowledge, only one French study has assessed the relationship between walkability 85 and cardiometabolic factors (body mass index (BMI) and blood pressure) (23). It was performed in the Paris metropolitan area - a highly dense urban setting – and used the Walk Score® (a commercial 86 instrument that is not yet supported by its developer in France, meaning that it is available but the 87 88 developer doesn't guaranty the validity in France) as the environmental indicator.

89

90 Given the differences in urban settings and behavioural patterns between Europe and 91 America, additional evidence is needed in each European country (26).

Hence, we analyzed data from the *Enquête Littorale Souffle Air Biologie Environnement* (ELISABET) survey of a representative sample of the population living in and around the cities of Lille and Dunkirk in northern France. The study's primary objective was to describe the association between neighbourhood walkability and cardiovascular risk factors among French adults. To this end,

96 we assessed a walkability index (WI), calculated using a geographic information system (GIS)) and the
97 Walk Score<sup>®</sup> (WS) as environmental indicators.

98

99 II. Methods

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- 101 1. Study population and area
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Our analysis was based on the cross-sectional ELISABET survey of a random sample for the electoral roll of men and women aged between 40 and 65 who have lived for at least on year in the Lille or Dunkirk urban areas, between January 2011 and November 2013. The ELISABET study's methodology has been described in detail elsewhere (27–29).

107

108 The ELISABET data were collected at home or in some case (10%) at the convenience of the 109 volunteer in a health care establishment; a trained, registered nurse administered a detailed 110 questionnaire, collected a blood sample and recorded anthropomorphic data. The study protocol was approved by the local investigational review board (CPP Nord Ouest IV, Lille, France; reference: 111 112 2010-A00065-34; ClinicalTrials.gov identifier: NCT02490553) in compliance with the French 113 legislation on biomedical research. All participants provided their written, informed consent to 114 participation in the study. Sampling frame and participation rate is described more extensively in 115 supplemental data

According to French census data, the Lille and Dunkirk urban areas respectively comprised 117 1,154,103 and 203,770 inhabitants in 2014 (French National Institute Statistics and Economics 118 Studies (*Institut National de la Statistique et des Études Économiques* (INSEE))(30). Each urban area 119 comprises an old city centre surrounded by agglomerated cities and non-agglomerated cities 120 separated by rural areas. leading to a wide range of walkability levels. Lille and Dunkirk both have a 121 dense public transport network, and traffic levels vary from low to high across each urban area.

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#### 2. Definition of cardiovascular risk factors

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- 125 2.1 Physical activity and walking
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Physical activity was evaluated using the International Physical Activity Questionnaire (31). 127 128 Three indicators were used to quantify physical activity: the intensity, and the frequency (per week) 129 and average duration (per day) for each intensity level. We defined three levels of physical activity, 130 according to the International Physical Activity Questionnaire results: high, moderate and low (see 131 supplemental material for the definition of the categories). We also analysed self-reported daily 132 walking time above 30 minutes as a binary variable (yes/no). For physical activity, we analysed the prevalence of moderate or high physical activity vs low and the prevalence of high activity versus low 133 or moderate. 134

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136 *2.2 Obesity* 

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We used the WHO definition of BMI, i.e. by the person's weight in kilograms divided by the square of his/her height in meters (kg/m<sup>2</sup>). Overweight corresponds to a BMI  $\geq$ 25 kg/m<sup>2</sup> and <30 kg/m<sup>2</sup>, and obesity corresponds to a BMI equal to or greater than 30 kg/m<sup>2</sup>. Height and weight were measured by a trained registered nurse during examination

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143 *2.3 Blood pressure* 

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145 Two consecutive measurements of systolic blood pressure (SBP) and diastolic blood pressure 146 (DBP) were made with participants in the sitting position, after 5 minutes of rest. Examination was 147 done in the morning, subject should be fasting for the appointment. When available, the blood pressure was the mean of two measurement. A second blood pressure measurement was taken systematically only during the last third of the study period (28) Hypertension was defined as an SBP ≥140 mmHg, a DBP ≥90 mmHg, or ongoing treatment with antihypertensive medication. To account for the effect of antihypertensive medication in our blood pressure analyses, we added 10 mmHg and 5 mmHg to the observed SBP and DBP values, respectively (32).

#### 153 2.4 Blood lipid profile and glycated haemoglobin (HbA1c) levels

154 The laboratory measurements of blood markers have been described previously (26). The fasting 155 HbA1c level in whole blood was measured using high performance liquid chromatography (VARIANT 156 II, Bio-Rad, Hercules, CA, USA). Serum cholesterol and TG levels were measured using enzymatic 157 assays. The level of high-density lipoprotein (HDL) cholesterol (HDL-C) was determined after the 158 precipitation of apolipoproteins B with phosphotungstate/magnesium chloride. The level of LDL-C 159 was calculated according to the Friedewald equation (LDL-C = Total Cholesterol - HDL-C - TG/5 in g/L). 160 Assay samples were collected after a 10-hour fast. Throughout the project, the quality of each 161 analysis was monitored in an internal quality control program (with the use of calibration standards, 162 laboratory blanks, and reference materials (Seronorm<sup>™</sup> Trace Elements Whole Blood, SERO, 163 Billingstad, Norway)) and an external quality control program (an interlab comparison program 164 established by the Quebec Toxicology Centre, Quebec National Institute of Public Health, Quebec, 165 Canada). All the samples in a given batch were analysed at the same time with the same calibration 166 standard. All biological samples were tested in the same laboratory.

167

168 *3. Exposure data* 

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170 Two frequently employed walkability measures (a WI and the WS) were used to assess171 neighbourhood walkability for each participant.

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#### 3.1 Walkability index

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The WI used here was based on the index developed by Frank and al(13, 27), as a function of net residential density, street connectivity, and land use diversity. To approximate walkable areas, a 500 m radius buffer zone was created around each participant's home address, using ArcGIS software (Esri, Redlands, CA, USA) and a 500m walkable distance from the address by the network for the second. These buffer zones were then used to collect data on the built environment accessible by foot for each participant. The main analysis was made with a 500 meters Euclidean buffer. Sensitivity analysis with a 500m network buffer were realised.

182 The net residential density corresponded to the number of residential dwellings (houses and 183 apartments) divided by the residential land surface within a participant's buffer zone. We used the 184 2010 population census data provided by the INSEE.

Street connectivity was obtained by calculating the intersection density. We used the 2013 road data from the French National Institute for Geographic and Forest Information (Institut National de l'Information Géographique et Forestière, BDTOPO® database). The ArcGIS software was first used to identify intersections of three or more walkable road segments. We excluded intersections on limited-access roads (e.g. motorways and flyovers). The intersection density was calculated by dividing the number of pedestrian-accessible street intersections in the buffer zone by the total area of the buffer zone.

Land use diversity corresponded to the types of destination that a participant could reach on foot in his/her neighbourhood. We considered retail facilities, institutions, amenities (e.g banks, post offices, hospitals, churches, etc.), recreational facilities, and residential facilities. To determine the number of destinations of each type within the buffer zone, we used data from the INSEE's 2016 Permanent Facilities Database (Base Permanente des Équipements). An entropy index was then calculated for each participant's buffer zone, using the following equation.

198 Entropy index of a buffer zone j = 
$$\frac{-\sum p_{ij} \times ln p_{ij}}{ln N_i}$$
 (34)

where p<sub>ij</sub> is the fraction of destinations of type i within the buffer zone j, and N<sub>i</sub> is the number of
different destination types (five, in the present analysis). The values range from 0 (a low diversity of
destinations) to 1 (high diversity).

Lastly, each indicator was standardized as a z-score, and the neighbourhood WI was obtained by summing the z-scores. The WI was considered as both a continuous variable (per one interquartile range (IQR) increment) and as a categorical variable (in tertiles).

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207 *3.2 The WS* 

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209 The WS was created by Front Seat Management LLC (Seattle, WA, USA). The algorithm 210 calculates neighbourhood walkability by assigning a one-mile radius (1.6 km – a 30 min walk) buffer 211 zone around the participant's residential address. Facilities present in the buffer zone are split into 212 seven categories, including dining & drinking facilities, grocery shops, other shops, places for errands, 213 parks, schools, and culture & entertainment. Based on the distance from the residential address (a 214 distance decay function), points are assigned for each destination in every category. No points are 215 given for destinations more than 1 mile away. The various (equally weighted) categories are summed 216 and normalized as a score ranging from 0 ("Car Dependent") to 100 ("Walker's Paradise"). The WS 217 also measures pedestrian friendliness by analysing the population density and road metrics such as 218 block length and intersection density. The WS's data sources include Google, education.com, 219 OpenStreetMap and places added by the WS user community.

In France, the WS is available but the score "is not yet supported in France", according to its developer; Front Seat Management LLC states that it does not have enough data to ensure an accurate score. However, a study of the Ile-de-France region (including Paris) found a strong association between objective walking time (measured with a GPS device and an accelerometer) and the neighbourhood WS (35). In the present study, each participants' residential addressed was

attributed with a WS corresponding to the nearest point on a 100 m grid extracted from the WSwebsite in January 2019.

227

228 4. Covariates

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In the present analysis, we selected participants who had lived at their current address for at
least one year. Participants with a non-localizable place of residence were excluded. The following
variables were recorded: age, sex, urban area (Lille or Dunkirk), educational level, professional status,
smoking status, marital status, and the identity of the nurse investigator having collected the study
data.

235 The median income and population density for each neighbourhood were extracted from 236 data provided by the French National Institute for Statistics and Economic Studies (INSEE). The 237 geographical unit for a neighbourhood corresponded to the French "regrouped statistical 238 information block" (IRIS) unit, as defined by the INSEE. The IRIS unit is the smallest census unit 239 available in France(36). We also recorded the annual mean residential PM<sub>10</sub> concentration, as 240 estimated by the ATMO Hauts-de-France monitoring organization with a spatial resolution of 25×25 241 m using an atmospheric dispersion modelling system. These methods have been described previously 242 (29).

243 For each outcome, we only analysed participants with a full corresponding dataset.

244

245 5. Statistical analyses

246

247 In order to describe the WI tertiles in each urban area, Chi<sup>2</sup> tests were used to compare 248 proportions of qualitative variables. Quantitative variables were assessed with Student's test (when 249 the data were normally distributed) or a Kruskal-Wallis test.

250 *5.1 Regression analyses* 

Geographical autocorrelation of quantitative outcome variables was estimated separately using Moran.I of the R package ape. All Moran indices were <0.02, therefore, auto-correlation were not addressed at the modelling stage.

254 To study the relationship between the outcomes (physical activity, BMI, blood pressure, 255 blood lipid markers, and HbA1c) and neighbourhood walkability, we used multiple linear regression 256 models for quantitative outcomes. We used Tobit regression model for blood lipids and HbA1c in 257 order to account for treatment effect. We used logistic regression models and robust Poisson 258 regression (glmrob of R Package robust) for binary outcomes. The WI and WS were used as 259 explanatory variables. The main model was adjusted for age (as a continuous variable), sex (as a 260 binary variable), urban area (Lille or Dunkirk), education level (primary education only, secondary 261 education only, 2-4 years of higher education, or 5 or more years of higher education), smoking 262 status (former smoker, never smoker, or current smoker), marital status (single, married or similar, 263 divorced, widower, or other), work activity (full-time work, part-time work, or unemployment/no 264 occupation), mobility limitation defined as answer yes to "Do you have difficulty to walk ?" or yes to 265 "Are you breathless when you walk with other people of your age on flat ground ?", investigator (a 266 class variable, with 12 modalities for the 12 nurse investigators), inclusion season (winter, summer, 267 fall and spring) and the median neighbourhood household income in the IRIS unit (as a continuous 268 variable). The variables are described in supplemental table 1.

We estimated mean differences for quantitative outcomes or the odds ratio for binary outcomes per tertile of neighbourhood walkability (using the WI and the WS, with the lowest tertile as reference) and for an interquartile range of neighbourhood walkability. We also tested for potential interactions between our main outcomes (BMI, SBP, and hypertension) on one hand and sex and urban area on the other. We tested interaction for age (years) and education (less than 2 years after high school / 2 years or more after high school) for the association between WI an

outcomes .For interaction, with p< 0.10, we presented stratified analysis (for the age the analysis</li>
were stratified by age superior or inferior to the median (53.5 years) )

277

#### 278 5.2 Sensitivity analyses

We calculated a WI with the most recent data available (2015 for resident density, 2020 for street connectivity and 2016 for permanent facilities) and calculated spearman correlation between the two scores in order to evaluate stability of WI over time.

282 Anti-hypertensive therapy is very common and may bias the results, standard adjustment is 283 not an accurate method for taking it in account. Many methods have been suggested to account the 284 treatment effect for blood pressure (37) . Conversely, few methodologies for treatment of 285 hyperlipemia or diabetes have been suggested. These various methodologies used for blood pressure 286 have strength and weakness. In order to evaluate the accuracy of the model and sensitivity to the 287 method, additional analyses were performed using three other methods. In the main model, we 288 added 10 mmHg to the SBP and 5 mmHg to DBP for treated participants (32). The first additional 289 method added a constant value to the participant's SBP and DBP, depending on the drug class and 290 the drug combination. We used the constant described in the ATOM study ("blood pressure-lowering 291 effects of antihypertensive drugs and combinations: meta-regression of published clinical trials") 292 (38,39). In the second method, we performed censored normal regression (using a Tobit model) with 293 right censoring on the observed blood pressure for a treated individual (37). Thirdly, we applied the 294 non-parametric algorithm described by Levy and al. (40).

Furthermore, we adjusted our analyses for the residential PM<sub>10</sub> and NO<sub>2</sub> concentration (as a have high concentrations of traffic-related air pollutants, which may raise the cardiovascular disease risk (29,41). Lastly, we performed mediation analysis (42), studying the mediator effect of physical activity (define as dichotomous variable low versus moderate or high) and BMI in the association between WI and blood pressure variables (SBP, DBP, and hypertension). We studied association

between physical activity and BMI with blood pressure and studied the ratio of mediation using the
 function mediate of the R package. The confident interval was calculated by Bootstrap with 500
 simulation.

303 We did further adjustment for household income and ethnicity. Missing data for house 304 income were code as category of income. Ethnicity (European or Other) were determined by cluster 305 the analysis of genetics data is currently ongoing (methodology not publish yet)(42)

306 In a sensitivity analysis, we take into account the sampling frame, using svydesign of the package

307 survey and the finite population correction to take all small town were not included in Lille city area.

308 To account the different sampling frame in Dunkirk and Lille, we did separate analysis in Lille and

309 Dunkerque and pooled the results using rma of the metaphor package. In this analysis, participants

310 reporting the use of diabetes medication or cholesterol medication were excluded from this

sensitivity analyses of the associations with HbA1c and the blood lipid profile, respectively.

312

313 The threshold for statistical significance was set to p<0.05. Statistical analyses were performed 314 using R software (version 3.5.1, R Core Team, 2018) (43)

315

316 III. <u>Results</u>

317

A total of 3276 participants had been included in the ELISABET survey, and one participant had been excluded after he withdrew his consent. The spatial distribution of the volunteers is presented in supplemental figure 1. In the present study, we excluded a further 43 participants because they had moved home within the previous year, 6 participants with missing data on their professional status, 4 with missing data on their marital status, 1 with missing data on his smoking status, and 1 who was unable to walk as the result of a known disease. Ultimately, 3218 participants were analyzed (1648 for Lille and 1570 for Dunkirk). Missing data meant that 125 participants were excluded from the physical activity analysis, 28 were excluded from the SBP analysis, and 56 wereexcluded from the analyses of DBP and hypertension.

327 The study population's characteristics are summarized by urban area and by WI level in Table 328 1. The participants' mean age was 53.3 yrs. In each urban area, participants living in the most 329 walkable neighbourhood were significantly more likely to be current smokers, single, or divorced, 330 and presented a higher educational level. In Lille, the most walkable neighbourhood had higher level 331 of PM<sub>10</sub> and a lower annual residential income; in Dunkirk, the opposite was true. The WI and WS 332 were closely correlated (Spearman's correlation coefficient = 0.77, p<1x10<sup>-15</sup>). The distribution of the 333 participants' WS values is described in Table 2; most of the participants lived in a "somewhat 334 walkable" or "very walkable" neighbourhood.

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#### Associations between neighbourhood walkability and cardiovascular risk factors

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338 The results of the multivariable analyses for each of the walkability indices are summarized in 339 Table 3. In the fully adjusted model, we observed a significant negative association between 340 neighbourhood walkability (according to the WI or the WS) and the prevalence of moderate or high 341 physical activity was increase by 12.3% [-0.1;26.2] in the high tertile of walkability. (OR for One IQR 342 [95%CI] = 1.25[1.13;1.38] ) and at least 30 minutes of walking per day was increased by 12.1% IC95% 343 [-2.2-28.5] ( OR 1.16[1.06;1.28] for a one IQR increment in WI). WS was associated to lower 344 prevalence of High physical activity -15.0%[-30.0;2.6] in the highest tertile of WS and OR 345 0.86[0.76;0.97] for one IQR. No significant association were observed with WI. There was a 346 significant association between neighbourhood walkability and BMI; for a one IQR increment in the WI, we observed a significant difference (- 0.23 [-0.44;-0.01] kg/m<sup>2</sup>) in the mean BMI. However, we 347 348 did not observe a significant association with obesity (OR 0.95[0.85;1.07] for WI). There were strong 349 negative associations between neighbourhood walkability and blood pressure outcomes. For a one 350 IQR increment in WI, there was a difference of - 1.66 [-2.46;-0.85] mmHg for the SBP and -0.79[-

1.31;-0.27] mmHg for the DBP. The prevalence of hypertension was -8.9%[-20.2;3.9] lower in the high tertile of walkability (OR for one IQR 0.87 [0.78;0.96]). We did not find significant associations with the biomarkers of cardiovascular risk (i.e. serum HbA1c, LDL-C, HDL-C and TG levels). All associations found for WI were also observed for WS. The only difference between the two scores was observed for a high level of physical activity.

356

357 Sensitivity analyses

We studied the association using network buffer of 500 m instead of the Euclidian buffer. Result were similar than in the main analysis (supplemental table 2). The spearman correlation between WI use in the main analysis and a WI calculated with the most recent data available was 0.95.

The results of the models for walkability and blood pressure after adjustment of the blood pressure values with four different approaches are summarized in Supplemental Tables 3a and 3b The associations with walkability and blood pressure were similar, although censored normal regression led to a slightly stronger association (-2.43 mmHg for the mean SBP for a one IQR increment in the WI).

The results of the models for neighbourhood walkability and cardiovascular risk factors after adjustment for the residential PM<sub>10</sub> or NO<sub>2</sub> concentration are summarized in Supplemental Table 3. The results were similar to those of the main model, except that the association was no longer significant for the BMI. No association between walkability index or Walkscore and Hbac1 was observed after adjustment for PM<sub>10</sub> Or NO<sub>2</sub>.

After adjustment for household income and divergent ancestry, results remain consistent (supplemental table 4). In the model taking into account the stratified sample, results were consistent except for the association of WS with physical activity and walking and WI with walking which were no more significant. Results were similar in the model taking into account the sampling

frame (Supplemental table 5) except for the association with physical activity which was only
significant for WI and moderate or high physical activity .

378 *Mediation analysis* 

379 Moderate or high physical activity was associated to lower diastolic blood pressure 380 compared to low physical activity (-0.98mmHG [-1.89; -0.08 mmHg]). No significant associations were observed between physical activity and blood systolic blood pressure or hypertension (both 381 382 p>0.6). Accordingly, association ratios between WI and blood pressure explain by physical activity 383 were small and non-significant (supplemental table 6). BMI was significantly associated to systolic 384 blood pressure (coefficient for 1 unit of BMI 0.91 mmHg [0.77; 1.04]), diastolic blood pressure ( 385 coefficient for 1 unit of BMI 0.57 mmHg [0.49 ; 0.63] and hypertension percentage increase or 386 prevalence 4.9% [3.9%; 5.9%]). Association ratios explained by BMI were significant and range from 387 12.1%[0.4%;27%] for systolic blood pressure to 17.4%[2.1%; 45%] for diastolic blood pressure.

388

#### 389 Interaction analysis

No significant interactions with the urban area or with sex were observed for any of the main outcomes (figure 1). For other interactions, the only significant one with WI was for education and daily walking above 30 minutes (p=0.04), the percentage increase in prevalence for one IQR of WI was 2.73 % [-5.04;11.62] in the lower education group and 9.42 % [0.19;20.16] in the higher education group. Interaction with p >0.10 are presented in supplemental table 7

### 395 Discussion

Our results showed that higher neighbourhood walkability was associated with a lower prevalence of low physical activity. Neighbourhood walkability had strong negative associations with blood pressure and BMI. However, we did not find a significant association with obesity or any of the blood markers.

400 With regard to blood pressure, the associations observed in the present study were 401 consistent with those reported previously in the United Kingdom(22) and in France (23). In the latter 402 study, the effect size for the association between WS and SBP (beta coefficient [95% CI] = -0.030 403 mmHg [-0.063;-0.0004] per unit increment in WS) was half that observed here (-0.064, [-0.095; -404 0.033]) (23). Our present results are also in agreement with data from North America(15,17,44) and 405 Australia (45). Consistency is found in the Bradford Hill criteria, which reinforces the hypothesis of 406 causality. Nevertheless these studies are mainly cross-sectional and further studies are needed to 407 conclude on causal association . This association may be explained by increased physical activity. In 408 the present study, neighbourhood walkability was negatively associated with a low physical activity 409 prevalence. However, no increase of high physical activity prevalence was observed, on the contrary, 410 a decrease in the prevalence with WS was observed. Walkability may increase moderate physical 411 activity through walking for daily activities. High physical activity, which depends on vigorous activity 412 (mainly accruing during sport practice), may be less impacted. Walkability may increase walking and 413 then reduce low physical activity but may not be sufficient to achieve high physical activity. 414 Consistently, sedentary behaviour is associated with high blood pressure (7), and moderate physical 415 activity has a positive impact on blood pressure (46,47). The activity was not associated to blood 416 pressure, accordingly, it does not explain the associations with SBP and DBP in the mediation 417 analysis. One reason for this small impact might be the misclassification of self-reported physical 418 activity and thus a lack of accuracy. Accordingly, previous studies did not clearly demonstrate 419 mediation by physical activity of the association between walkability and changes in cardiometabolic 420 outcomes (45,48,49). The small impact of physical activity on the association might also be due to 421 one or more other mechanisms or cofounding factors acting alone or together. Future research 422 should investigate additional mechanisms by which walkability is involved in cardiometabolic health 423 outcomes.

424 With regard to weight status, we observed a significant association with BMI but not with 425 obesity. The negative association with BMI was consistent with European studies (conducted in

426 Belgium (50) and France (23)) and American studies (51,52). The effect size reported in the French 427 study (estimated  $\beta$  coefficient [95% CI] for BMI = -0.010 kg/m<sup>-2</sup> [-0.019;-0.002] per unit increment in 428 WS (23)) was similar to that observed here (-0.011 [-0.020; -0.002]). The earlier French study also 429 found that living in a low-walkability (car-dependent) neighbourhood was associated with an 430 elevated prevalence of obesity. Furthermore, longitudinal studies in North America found a strong, 431 protective association between higher walkability and obesity (15). In contrast, we did not find a 432 significant association with obesity in our study perhaps because of loss of power analysing a 433 binomial variable (obesity) instead of a continuous variable (BMI).

Lastly, the associations between neighbourhood walkability and SBP and DBP were only partly explained in the mediation analysis by BMI. The effect of walkability on blood pressure through a lowering in BMI does not seems to be predominant - probably due to a small weight reduction. Even though weight reduction diminishes blood pressure (53), other mechanisms may be involved, previous studies suggest that physical activity may reduce blood pressure through physiological pathway other than body weight (54).

440 With regard to blood markers, we did not find an association with HbA1c and even a non-441 significant increase of HbA1c associated with high walkability in the main model was observed. A 442 recent meta-analysis reported that neighbourhood walkability was associated with a lower 443 risk/prevalence of diabetes (55). Two European studies did not find an association between 444 walkability and type 2 diabetes. The effect of walkability might simply not be strong enough to have 445 any observable effects on the HbAc1 level. Furthermore, HbA1c has been linked to air pollution. As 446 regards, one possible explanation for the lack of association of HbA1c with high walkability in the 447 main model could be due to higher air pollution in area with high density. Accordingly, air pollution 448 was higher in walkable neighbourhoods in Lille. The non-significant, unfavourable association 449 between walkability and HbA1c disappeared after adjustment for PM<sub>10</sub> or NO<sub>2</sub>. Although it is not 450 sufficient to confirm it this last result is consistence of the hypothesis of a confusion by air pollution.

451 In sedentary people, increased light- or moderate-intensity physical activity has a positive effect on blood lipid levels by increasing the HDL-C level and decreasing the LDL-C and triglyceride 452 453 levels (56–58). However, we did not find an association between neighbourhood walkability and any 454 of the blood lipid markers. Again, the influence of walkability on physical activity might not be strong 455 enough to produce an observable effect on the blood lipid profile. Canadians studies have observed than higher HDL cholesterol was associated to Walkability(16,18). However, these results are 456 457 inconsistent to the best of our knowledge, no European studies have been conducted, and an 458 American study did not observe any association with HDL-C or LDL-C (17). Furthermore, a recent 459 Dutch study did not find any evidence to suggest that population density (a walkability parameter) is 460 associated with blood lipid levels (50). Therefore, there is limited evidence for an association 461 between walkability and blood lipids.

We observed a higher prevalence of smoking in area with high walkability. This association has been observed previously in Ontario. Nevertheless, we didn't expected walkability to have unfavourable impact on smoking. The reasons for higher rate of smoking associated to walkability remain to be explained. One hypothesis may be the greater proximity to tobacco seller(16)

466 Our results were consistent across cities, gender, education groups and age. A significant 467 interaction was observed only for walking ≥ 30 min/Day and education. It seems that walkability may 468 be more associated to walking when education level is higher. This may be due to a possible increase 469 in health consciousness among better educated people. Il may also be due to chance due to the large 470 number of interaction tests done.

471

Our study had some limitations. Firstly, its cross-sectional design prevented us from assessing the possible causal nature of the relationship between neighbourhood walkability and our outcomes. Reverse causation might be possible due to potential self-selection of neighborhood driven by physical activity preferences. Health-conscious participants may selected residential location that facilitated physical activity; in turn, that choice might lead to overestimation of the relationships

between walkability and health outcomes(59). In our study, however, residents living in more
walkable areas reported smoking more (i.e. arguing against this hypothesis), and previous studies
have shown that reverse causation is unlikely (60,61).

480 Secondly, there was a time interval between the collection of the ELISABET data (in 2011-2013) and 481 the collection of geographical data (in 2019 for the WS and from 2013 to 2016 for the WI). It is 482 possible that some neighbourhoods changed during this time, which might have led to 483 misclassification of walkability measures. Nevertheless, any changes might have been marginal. 484 Moreover, we did not account for fine features such as greenness, sidewalk conditions, or shade – all 485 of which can influencing pedestrian behaviours like leisure-time walking (56) accordingly the 486 correlation between the two WI calculated with different data periods were high. Thirdly, we could 487 not incorporate more subjective variables, such as perceived security or aesthetics. Fourthly, the 488 participant individual income measurement had weakness, there were lots of missing data and only 489 household income was available. Residual confounding on socio economic level is therefore possible. 490 Furthermore, we didn't have ethnicity information. Nevertheless, results were consistent after 491 adjusting with a proxy of ethnicity by genetic cluster analysis which is not validated yet.

492 Filthy, This work has an analytic objective and representativeness of the sample is a less 493 important issues(63). Nevertheless, we take into account the study sampling in a sensitivity analysis. 494 The results on blood pressure remain significant. Association with physical activity were consistent 495 but only significant for WI. The reduce number of significant results with physical activity may be 496 explain by the lower power of this analysis. Indeed the number of degrees of freedom were increase 497 by the stratification of the analysis by city areas. Lastly, our walkability index did not distinguish 498 between establishments within each type of destination; for example, the retail category included 499 supermarkets and small convenience stores. Nevertheless, this measurement error might have been 500 non-differential and would have biased the association towards the null

501 This study had several strengths. Firstly, we used objective measures of cardiometabolic risk markers.
502 Secondly, we use two different walkability indices - one of which was GIS-based. Thirdly, we studied

two different French cities with a wide range of urban settings and environmental features. The observed associations were similar in both cities. Furthermore, the observed associations were consistent with both walkability indices and all methods of adjusting the blood pressure analysis. This strengthens the level of confidence in our associations - especially for blood pressure and BMI, which do not appear to be explained by specific local features or methodological issues.

- 508
- 509
- 510 Conclusion
- 511

512 Overall, our results evidenced favourable associations between neighbourhood walkability on one 513 hand and blood pressure and BMI on the other. However, no association was found for other 514 cardiovascular health outcomes (HbA1c and blood lipid markers) in the French context. Greater 515 neighbourhood walkability might improve health at the population level. Further research in Europe 516 is required for a better understanding of the mechanisms involved - especially for blood markers.

517

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535

### 536 Competing interest

537 AC, FO, PA, AM declare that they have no conflicts of interest.

538

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- 735

### 737 Figure 1. Association with main outcomes for an IQR increment in the WI



		Lille				Dunkirk		
	Low WI	Medium WI	High WI		Low WI	Medium WI	High WI	
Individual Variables				P-Value <sup>a</sup>				P-Value <sup>a</sup>
Number n(%)	495(30.0)	592(35.9)	561(34.0)		584(37.2)	483(30.8)	503(32.0)	
Sex (Men), n(%)	240(48.5)	275(46.5)	255(45.5)	0.61	282(48.3)	246(50.9)	236(46.9)	0.44
Age, mean (SD)	53.2(6.9)	53.5(7.2)	52.6(7.4)	0.11	53.4(7.3)	53.4(7.4)	53.3(7.2)	0.942
Educational level, n(%)				<10 <sup>-4</sup>				<10 <sup>-4</sup>
5 years or more after high school	99(20)	135(22.8)	153(27.3)		58(9.9)	50(10.3)	91(18.1)	
2 years to 4 years after high school	118(23.8)	131(22.1)	114(20.3)		87(14.9)	65(13.5)	75(14.9)	
secondary education	242(48.8)	277(46.7)	239(42.6)		373(63.9)	299(61.9)	275(54.7)	
primary education	36(7.2)	49(8.3)	55(9.8)		66(11.3)	69(14.3)	62(12.3)	
Marital status, n(%)				<10 <sup>-4</sup>				<10 <sup>-4</sup>
single	25(5.0)	47(7.9)	93(16.6)		12(2.0)	24(5)	40(7.9)	
married or similar	405(81.8)	449(75.84)	363(64.7)		526(90.1)	393(81.4)	373(74.1)	
divorced	49(9.9)	70(11.82)	91(16.2)		34(5.8)	53(11)	71(14.1)	
widowed	12(2.4)	22(3.7)	10(1.8)		12(2.0)	13(2.7)	18(3.6)	
other	4(0.8)	4(0.7)	4(0.7)		0(0)	0(0)	1(0.2)	
Work activity, n(%)	151/21 1	100/01 0		0.733	225110.0		105/04 0	0.164
no occupation/unemployed	154(31.1)	188(31.8)	178(31.7)		236(40.4)	215(44.5)	185(36.8)	
full time work	30(11.3) 285(57.6)	/8(13.2)	/8(13.9)		61(10.5)	47(9.7)	61(12.1)	
Spectrum status $p(%)$	285(57.6)	320(33.1)	505(54.4)	0.01	287(49.1)	221(45.8)	237(31.1)	0 186
Nonsmokers	582(53.6)	556(49.6)	517(49)	0.01	320(54.8)	246(50.9)	241(47.9)	0.180
Current smokers	169(15.6)	217(19.4)	226(21.4)		97(16.6)	240(30.9)	107(21.3)	
Former smokers	335(30.8)	347(31)	312(29.6)		167(28.6)	147(30.4)	155(30.8)	
Physical activity layed $p(%)^{b}$	555(50.0)	517(51)	512(25.0)	< 10 <sup>-4</sup>	107(20.0)	117(50.1)	155(50.0)	0.003
Low	144(29.7)	149(25.9)	156(28.1)	<b>NIO</b>	301(51.4)	217(54.4)	203(42.5)	0.005
Moderate	142(29.3)	249(43.3)	241(44.1)		165(28.2)	159(29.8)	192(40.1)	
High	198(41)	177(30.7)	150(27.4)		87(14.1)	80(15.7)	83(17 4)	
Daily walking >30 min, $n(\%)$	210(42.42)	222(37.5)	251(44.74)	0.036	223(38.18)	193(39.96)	221(43.94)	0.141
BMI (kg/m <sup>2</sup> ), mean (SD)	26.9(4.7)	26.6(5.2)	26.6(5.3)	0.553	27.5(5.1)	28(5.2)	27.1(5.2)	0.039
Weight status $n(\%)$		· · /	× /	<10 <sup>-4</sup>	· · · ·			0.025
BMI<25	202(41)	247(41.7)	239(42.6)	10	205(35.1)	142(29.4)	184(36.6)	0.020
25 <bmi<25< td=""><td>178(36)</td><td>233(39.3)</td><td>205(36.5)</td><td></td><td>236(40.4)</td><td>187(38.7)</td><td>195(38.8)</td><td></td></bmi<25<>	178(36)	233(39.3)	205(36.5)		236(40.4)	187(38.7)	195(38.8)	
BMI>30	115(23)	112(19)	117(20.9)		143(24.5)	154(31.9)	124(24.6)	
SBP (mmHg) mean (SD) <sup>c</sup>	129.2 (19.3)	126 5 (19)	126 5 (18 2)	0.035	132 1(19 3)	130 (19)	128 2(20)	0.039
$DPP (mmHg) maan (SD)^d$	82 2(11.6)	81 8(11.5)	82 5(11 5)	0.158	82 7(12 2)	82 7(12 1)	81.0(12.2)	0.022
UDD ((1) <sup>d</sup>	03.2(11.0)	01.0(11.3)	02.3(11.3)	0.156	000(40.7)	03.7(12.1) 025(49.6)	81.9(12.2) 220(42.7)	0.022
HBP, n(%)	231(46.7)	253(42.7)	237(42.2)	.306	290(49.7)	235(48.6)	220(43.7)	0.132
Antihypertensive medication, $n(\%)$	119(24)	142(24)	129(23)	0.899	145(24.8)	119(24.6)	118(25.5)	0.850
LDL-C'(g/L), mean (SD)	1.41(0.32)	1.37(0.36)	1.4(0.36)	0.151	1.41(0.3)	1.40(0.35)	1.42(0.34)	0.811
HDL- $C^{e}$ (g/L), mean (SD)	0.6(0.1)	0.59(0.1)	0.58(0.1)	0.17	0.58(0.14)	0.56(0.14)	0.57(0.15)	0.08
$HbA1c^{T}(\%)$ , mean (SD)	5.58(0.6)	5.65(0.6)	5.74(0.7)	<10 <sup>-4</sup>	5.75(0.67)	5.78(0.76)	5.78(0.76)	0.772
Environmental variables								
Walk Score® mean (SD)	39 2(22 7)	60 4(18 7)	79 7(13)	< 10 <sup>-4</sup>	33 7(24 2)	58 2(22 2)	76 8(20 8)	< 10 <sup>-4</sup>
IRIS median income	21.1 [19.8-22.4]	20.6[18.1-21.5]	17.2[13.4-21.6]	<10 <sup>-4</sup>	17 [16.4-20]	16.7[15.5-18.9]	18.1[14.1-20]	<10 <sup>-4</sup>
(×1000 euros), median [IQR] IRIS density (1000 hab/km 2), median [IOR]	1.6[0.6-3.3]	4.0[1.8-5.1]	8.5[6.0-11.3]	<10 <sup>-4</sup>	0.9[0.4-2.5]	4.1[2.3-6.3]	6.9[4.5-9.1]	<10 <sup>-4</sup>
$PM10 (\mu g/m 3) mean(sd)$	25.9(1.8)	26.9(1.8)	28.15(1.6)	<10 <sup>-4</sup>	26.2(1.1)	26.63(0.8)	25,19(0.6)	<10 <sup>-4</sup>
NO2 (ug/m 3) maan(sd)	22.2(4.0)	25.0(4.7)	28 72(4 8)	<10 <sup>-4</sup>	19 055(2 19)	20 222(2 2)	20 826(1 72)	~10_2*
$102 (\mu g/m 5), mean(su)$	23.2(4.9)	23.9(4.7)	28.12(4.8)	N10	19.000(2.10)	20.222(2.2)	20.020(1.12)	<10-3

Table 1 : Participants included in the analysis and neighborhood characteristics, by urban area and level of walkability

WI: walkability index; n: number; SD: standard deviation.

<sup>a</sup> Student's t-test for quantitative variables (Kruskal-Wallis test for skewed quantitative variables), or a chi<sup>2</sup> test for qualitative variables

740b: 125 missing data, c : 28 missing data; d : 56 missing data

	Table 1 (continuing) : Participants	included in the analysis and neighborhood characteristics, by urban area and level of walkability	
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		Lille				Dunkirk		
	Low WI	Medium WI	High WI		Low WI	Medium WI	High WI	
Individual Variables				P-Value <sup>a</sup>				P-Value <sup>a</sup>
Genotype clusters ,n(%)								0.317
European	444(89.7)	539(91.05)	464(82.71)	<10 <sup>-4</sup>	532(91.1)	427(88.41)	448(89.07)	
Others	51(10.3)	53(8.95)	97(17.29)		52(8.9)	56(11.59)	55(10.93)	
Inclusion season ,n(%)				0.726				0.622
Autumn	111(22.42)	137(23.14)	141(25.13)		150(25.68)	141(29.19)	123(24.45)	
Summer	104(21.01)	140(23.65)	122(21.75)		128(21.92)	104(21.53)	110(21.87)	
Winter	129(26.06)	132(22.3)	132(23.53)		134(22.95)	110(22.77)	112(22.27)	
Spring	151(30.51)	183(30.91)	166(29.59)		172(29.45)	128(26.5)	158(31.41)	
MET,median [IQR]	2235[887.1-	1806[856.5-	1(2(1702 22(0)	4 0-4		1039.5[280.1-	1227.8[462-	0 111
	5040]	3724]	1626[/92-3360]	<10	996[330-2346]	2529.75]	2362.75]	0.111
Annual Household income				·10 <sup>-4</sup>				
(euros),n(%)				<10				0.036
<7000	124(25.05)	165(27.87)	158(28.16)		96(16.44)	99(20.5)	94(18.69)	
7000-15 000	4(0.81)	3(0.51)	13(2.32)		1(0.17)	1(0.21)	3(0.6)	
15 000-30 000	21(4.24)	42(7.09)	71(12.66)		15(2.57)	26(5.38)	28(5.57)	
30 000-45 000	142(28.69)	168(28.38)	154(27.45)		116(19.86)	69(14.29)	73(14.51)	
45 000-60 000	101(20.4)	109(18.41)	67(11.94)		47(8.05)	45(9.32)	41(8.15)	0.164
60 000	54(10.91)	57(9.63)	51(9.09)		29(4.97)	25(5.18)	32(6.36)	
Didn't answer the question.	49(9.9)	48(8.11)	47(8.38)		280(47.95)	218(45.13)	232(46.12)	
Mobility limitations, n(%)	36(7.27)	39(6.59)	57(10.16)	0.064	72(12.33)	60(12.42)	52(10.34)	0.505
Mobility limitations, n(%)	49(9.9) 36(7.27)	46(8.11) 39(6.59)	57(10.16)	0.064	72(12.33)	60(12.42)	52(10.34)	0.5

741 MET: Metabolic Equivalent of Task

	Lill	e	Dunk	nkirk	
Walk Score®	n(%)	Mean	n(%)	Mean	
Overall	1648	60.6	1570	55,1	
Very car-dependent (0-24)	201(12.2)	14.7	373(23,8)	11.1	
Car dependent (25-49)	215(13.0)	34.1	159(10,1)	40.6	
Somewhat walkable (50 -69)	504(30.6)	60.2	418(26,6)	59.0	
Very walkable (70 -89)	605(36,7)	78.8	459(29,2)	78.6	
Walker's paradise (90 -100)	123(7,5)	95.1	161(10,3)	93.8	

Table 2 : Distribution of the Walk Score® by urban area

Variables				% of increa	ase		Odds ratio			D	ifference of M ean	
			Tertile of walka	ability	Linear trend for an interquartile		Odds ratio for an interquartile		Tertile of wal	kability	Linear trend for an interquartile	
	Score	Low (ref)	Medium [95% CI]	High [95% CI]	range increment in walkability [95%	CI] p-Value	Linear trend for an interquartile	p-Value	Low (ref Medium [95% CI]	High [95% CI]	range increment in walkability [95%	p-Value
Physical activity												
Moderate or High activity <sup>b</sup>	WI	-	9.49[-2.16;22.52]	12.26[-0.12;26.16]	6.9[1.2;12.72]	0.01744	1.25[1.13;1.38]	0.00001				
	WS	-	4.49[-6.78;17.12]	9.49[-2.49;22.94]	5.14[-1.06;11.36]	0.10413	1.17[1.06;1.3]	0.00258				
High activity <sup>b</sup>	WI	-	-11.11[-25.24;5.69]	-12.67[-27.62;5.37]	-5.29[-14.52;4.28]	0.27489	0.94[0.83;1.05]	0.26169				
	WS	-	-7.12[-22.03;10.65]	-15.02[-29.59;2.56]	-10.8[-20.25;-1.31]	0.02572	0.86[0.76;0.97]	0.01087				
Daily Walking≥ 30 min	WI	-	-4.34[-16.5;9.6]	12.1[-2.22;28.51]	7.91[1.07;14.92]	0.02306	1.16[1.06;1.28]	0.00131				
	WS	-	-1.82[-14.36;12.56]	8.59[-5.36;24.59]	4.79[-2.6;12.2]	0.20438	1.1[0.99;1.21]	0.07125				
MET(%)	WI								15.84[-32.58;5.06]	11.73[-21.1;58.21]	11.51[1.99;21.37]	0.01748
	WS								11.52[-27.27;7.66]	-7.32[-28.11;19.49]	4.37[-5.54;14.32]	0.38826
Weight status												
BMI (kg/m <sup>2</sup> )	WI								0.03[-0.44;0.39]	-0.3[-0.73;0.13]	-0.23[-0.44;-0.01]	0.03723
	WS								0.19[-0.61;0.23]	-0.32[-0.75;0.1]	-0.26[-0.48;-0.03]	0.02593
Obese	WI	-	5.04[-11.91;25.26]	-5.38[-21.7;14.35]	-3.5[-12.73;6.07]	0.4687	0.95[0.85;1.07]	0.40163				
	WS	-	2.68[-13.92;22.49]	-11.64[-26.97;6.91]	-5.87[-15.72;4.01]	0.24367	0.92[0.82;1.03]	0.15616				
Blood pressure												
SBP (mmHg) <sup>c</sup>	WI								2.81[-4.39;-1.24]	-2.7[-4.33;-1.07]	-1.66[-2.46;-0.85]	0.00006
	WS								2.09[-3.69;-0.5]	-2.28[-3.91;-0.65]	-1.52[-2.38;-0.66]	0.00054
DBP (mmHg) <sup>d</sup>	WI								0.92[-1.94;0.1]	-1.1[-2.15;-0.04]	-0.79[-1.31;-0.27]	0.0031
	WS								0.77[-1.81;0.26]	-0.89[-1.95;0.16]	-0.69[-1.25;-0.14]	0.01461
Hypertension <sup>d</sup>	WI	-	-6.98[-17.95;5.46]	-9.42[-20.64;3.39]	-7.12[-13.56;-0.52]	0.03461	0.86[0.78;0.95]	0.00225				
	WS	-	-5.93[-17.18;6.85]	-8.9[-20.16;3.94]	-6.69[-13.56;0.2]	0.05719	0.87[0.78;0.96]	0.00635				
Current smoker	WI	-	2.93[-16.88;27.47]	19.51[-3.41;47.88]	12.48[1.91;23.46]	0.02026	1.2[1.07;1.35]	0.00201				
	WS	-	24[-0.35;54.3]	40.22[12.78;74.33]	21[8.86;33.2]	0.00068	1.3[1.14;1.49]	0.0001				
Hba1c (%)	WI								- 0.03[-0.03;0.08]	0.07[0.01;0.13]	0.02[-0.01;0.05]	0.15841
()	WS								- 0.02[-0.04;0.08]	0.05[-0.01;0.11]	0[-0.03;0.04]	0.79696
Lipidemia												
LDL(g/L)	WI								0.01[-0.02;0.01]	0[-0.02;0.01]	0[-0.01;0.01]	0.82986
(())	WS								0.01[-0.02;0.01]	0[-0.02;0.01]	-0.01[-0.01;0]	0.10571
HDL(g/L)	WI								0.01[-0.04;0.02]	0[-0.03;0.03]	-0.01[-0.02;0.01]	0.43311
(0) )	WS								0.03[-0.06;0]	-0.02[-0.05;0.01]	-0.01[-0.03;0.01]	0.29618
TG (g/L)f	WI								0.02[-0.06;0.01]	-0.01[-0.04;0.03]	-0.01[-0.02;0.01]	0.45793
	WS								0.04[-0.08:-0.01]	-0.03[-0.07:0.01]	-0.01[-0.03:0.01]	0.16389

CI: confidence interval; IQR: interquartile range; MET: Metabolic Equivalent of Task

main model adjusted for sex, age, urban area, educational level, smoking status, inclusion season, marital status, work activity, mobility limitation and residential median income

b: 125 missing data, c: 28 missing data; d: 56 missing data

746 Mutiple linear regression for difference of mean; poisson regression for % of increase; logistice regression for odds ratio

Walkability can be defined as the extent to which the built environment supports and encourages walking trips We defined walkability using three parameters:

household density street connectivity the land-use mix





We analyzed associations between walkability and cardiovascular risk factor

For a one interquartile increment in walkablity :



the % of prevalence increase was 5,8% for daily walking above 30 minutes



the BMI fell by 0.23 kg.m<sup>-2</sup>



the systolic blood pressure fell by 1.66 mmHg