



Objectively Measured Physical Activity and Sedentary Time Are Associated With Cardiometabolic Risk Factors in Adults With Prediabetes: The PREVIEW Study

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OBJECTIVE

The aim of the present cross-sectional study was to examine the association among physical activity (PA), sedentary time (ST), and cardiometabolic risk in adults with prediabetes.

RESEARCH DESIGN AND METHODS

Participants ($n = 2,326$; 25–70 years old, 67% female) from eight countries, with a BMI $>25 \text{ kg} \cdot \text{m}^{-2}$ and impaired fasting glucose ($5.6\text{--}6.9 \text{ mmol} \cdot \text{L}^{-1}$) or impaired glucose tolerance ($7.8\text{--}11.0 \text{ mmol} \cdot \text{L}^{-1}$ at 2 h), participated. Seven-day accelerometry objectively assessed PA levels and ST.

RESULTS

Multiple linear regression revealed that moderate-to-vigorous PA (MVPA) was negatively associated with HOMA of insulin resistance (HOMA-IR) (standardized $\beta = -0.078$ [95% CI $-0.128, -0.027$]), waist circumference (WC) ($\beta = -0.177$ [$-0.122, -0.134$]), fasting insulin ($\beta = -0.115$ [$-0.158, -0.072$]), 2-h glucose ($\beta = -0.069$ [$-0.112, -0.025$]), triglycerides ($\beta = -0.091$ [$-0.138, -0.044$]), and CRP ($\beta = -0.086$ [$-0.127, -0.045$]). ST was positively associated with HOMA-IR ($\beta = 0.175$ [$0.114, 0.236$]), WC ($\beta = 0.215$ [$0.026, 0.131$]), fasting insulin ($\beta = 0.155$ [$0.092, 0.219$]), triglycerides ($\beta = 0.106$ [$0.052, 0.16$]), CRP ($\beta = 0.106$ [$0.39, 0.172$]), systolic blood pressure (BP) ($\beta = 0.078$ [$0.026, 0.131$]), and diastolic BP ($\beta = 0.106$ [$0.39, -0.172$]). Associations reported between total PA (counts $\cdot \text{min}^{-1}$), and all risk factors were comparable or stronger than for MVPA: HOMA-IR ($\beta = -0.151$ [$-0.194, -0.107$]), WC ($\beta = -0.179$ [$-0.224, -0.134$]), fasting insulin ($\beta = -0.139$ [$-0.183, -0.096$]), 2-h glucose ($\beta = -0.088$ [$-0.131, -0.045$]), triglycerides ($\beta = -0.117$ [$-0.162, -0.071$]), and CRP ($\beta = -0.104$ [$-0.146, -0.062$]).

CONCLUSIONS

In adults with prediabetes, objectively measured PA and ST were associated with cardiometabolic risk markers. Total PA was at least as strongly associated with cardiometabolic risk markers as MVPA, which may imply that the accumulation of total PA over the day is as important as achieving the intensity of MVPA.

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The global prevalence of diabetes among adults has almost quadrupled since 1980 to 422 million cases in 2014, and continued growth is expected (1). This dramatic rise in prevalence is largely due to the increase in type 2 diabetes, which accounts for the majority of all diagnosed cases in adults (1). Changes in lifestyle factors such as sedentary behavior, insufficient physical activity (PA), dietary choices, and excess weight are important contributors to the development of type 2 diabetes (2). PA plays an important role, independent of weight, in the prevention of type 2 diabetes through its effect on insulin resistance (3). PA leads to the translocation of GLUT4 transporters to the plasma membrane, increasing glucose uptake into skeletal muscle (4). Sedentary time (ST) has also been associated with insulin resistance and fasting triglyceride levels, independent of PA and obesity (5). ST is thought to affect carbohydrate metabolism by decreasing muscle GLUT4 concentrations while also reducing lipoprotein lipase activity and triglyceride clearance (6,7).

Light activity is also associated with cardiometabolic health and, together with ST, occupies the majority of waking hours (8). Indeed, light-intensity activity substantially contributes to overall daily energy expenditure and may also mean spending less time in sedentary behaviors.

Many studies reporting associations between PA, ST, and cardiometabolic risk factors are limited by the self-reported measures of PA and ST (9,10), which are susceptible to reporting and recall bias. In the limited number that have used objective measures of PA, investigations of international samples measured concurrently with the same protocol are lacking and little consistency exists with regard to the devices, wear-time criteria, intensity cut points, or epoch lengths used, thereby limiting the ability to make comparisons between these studies. Furthermore, despite those with prediabetes being at the highest risk of developing type 2 diabetes and representing the population for which many lifestyle interventions are targeted, few studies have described the relationship between PA and insulin resistance in this population.

Typically, epidemiological research assessing the effect of PA and ST on cardiometabolic risk factors has been conducted in the general population (8,11,12). The few studies conducted in high-risk populations

have included participants based on a risk score questionnaire (5) or family history of type 2 diabetes (9,13), or participants diagnosed with diabetes and prediabetes have been combined (14). Consequently, physical characteristics and metabolic parameters of study participants varied substantially. As a result, it is unclear to what extent reported associations could be inferred to individuals with prediabetes.

Therefore, the purpose of this study was to quantify the relationship between objectively measured PA and ST with cardiometabolic health and risk of diabetes.

RESEARCH DESIGN AND METHODS

Participants and Setting

A detailed account of the PREvention of diabetes through lifestyle Intervention and population studies in Europe and around the World (PREVIEW) project has previously been published (15). Participants were recruited into PREVIEW between June 2013 and February 2015. PREVIEW is a large multinational diabetes prevention intervention being conducted at eight study sites: University of Copenhagen (Denmark), University of Helsinki (Finland), University of Maastricht (the Netherlands), University of Nottingham (U.K.), University of Navarra (Spain), Medical University of Sofia (Bulgaria), University of Sydney (Australia), and University of Auckland (New Zealand).

Participants were selected through an Internet-based prescreening tool or telephone interview using the Finnish Diabetes Risk Score (16). A total of 15,611 individuals were contacted for prescreening. Potential participants were sent a written description of the trial, were given verbal information at the study site, and signed informed consent prior to a laboratory screening. The laboratory screening was attended by 5,472 participants and included assessment of body mass, stature, and resting blood pressure (BP) and a 2-h oral glucose tolerance test (OGTT) (17). Glucose concentrations were analyzed at each study site (HemoCue, Angelholm, Sweden; Reflotron, Roche Diagnostics, Switzerland; or EML105 Radiometer, Copenhagen, Denmark) to identify people with preexisting diabetes.

At the end of this process, 2,326 participants met the following inclusion criteria and were found eligible to take part in the study: age 25–70 years, BMI > 25 kg/m², and prediabetes confirmed at OGTT. Prediabetes

was defined in line with World Health Organization/International Diabetes Federation and American Diabetes Association criteria (17) as either 1) impaired fasting glucose, with venous plasma glucose concentration of 5.6–6.9 mmol · L⁻¹, or 2) impaired glucose tolerance, with venous plasma glucose concentration of 7.8–11.0 mmol · L⁻¹ at 2 h and fasting plasma glucose < 7.0 mmol · L⁻¹. Participants were free of preexisting type 2 diabetes and of any illness and/or medication with a potential effect on compliance or the outcomes of the study.

Measurements and Procedures

Physical Activity

Participants wore an ActiSleep+ (ActiGraph LLC, Pensacola, FL) accelerometer attached to an elastic waist belt worn over the right midaxillary line. The ActiSleep+ was worn 24 h · day⁻¹ for seven consecutive days, with removal only for water-based activities. The principal output from the ActiSleep+ is an activity count, which represents raw accelerations that have been rescaled and filtered. Activity counts were collected at 100 Hz and aggregated to 60-s epochs (18). Sleep time was determined using a fully automated algorithm developed for use in 24-h waist-worn accelerometer protocols. The algorithm produces estimates of a nocturnal sleep period that are compared with an expert visual inspection of accelerometer trace (19). After the removal of nocturnal sleep episodes, participants were included in the analyses if they wore the monitor for ≥ 10 h on ≥ 4 days (18) including 1 weekend day (20). Mean activity count during valid wear time (counts · min⁻¹ [CPM]) has been shown to correlate well with total activity energy expenditure measured by the doubly labeled water technique (21) and was used as an indicator of total PA volume (18,22). Troiano cut points (18) were used to determine time (min · day⁻¹) spent at different intensity categories (sedentary < 100, moderate < 2,020, and vigorous < 5,999 CPM). Moderate and vigorous activity were summed to obtain moderate-to-vigorous PA (MVPA).

Cardiometabolic Risk Factors

In all study centers, standardized procedures were followed and measurements were performed by trained personnel. Self-administered questionnaires, accelerometer data, and fasting blood samples were collected at baseline. Blood was

drawn from the vein in the antecubital fossa after fasting (>10 h). Blood samples were initially stored locally at -80°C and then transported and analyzed centrally at the National Institution for Health and Welfare (THL) in Helsinki, Finland, where they were analyzed for insulin, HbA_{1c} , glucose, hs-CRP, total cholesterol, triglycerides, HDL cholesterol, and LDL cholesterol concentrations. Insulin resistance was calculated using HOMA of insulin resistance (HOMA-IR) with the following equation: $\text{HOMA-IR} = \text{fasting insulin (mU} \cdot \text{L}^{-1}) \times \text{fasting glucose (mmol} \cdot \text{L}^{-1}) / 22.5$. HOMA-IR has been validated against the gold standard hyperinsulinemic-euglycemic clamp technique (23). Total adiposity was assessed by DEXA at four sites (Copenhagen, Nottingham, Sydney, and Auckland), by bioelectrical impedance at three (Helsinki, Sofia, and Navarra), and with BOD POD (Maastricht) at one. Self-administered questionnaires assessed general and socioeconomic variables, including ethnicity, educational status, and household income.

Statistical Analysis

Descriptive statistics (mean \pm SD) were calculated for continuous variables and frequencies (%) for categorical variables.

Two-fifths of the participants had a missing value on at least one variable; HOMA-IR was missing in 7.8% of all cases, while 17% of values were missing for the accelerometer values (CPM, MVPA, and ST). Multiple imputation with a fully conditional specification model (Markov chain Monte Carlo) was used to impute missing values. All variables were included in the imputation, and all variables with skewed distribution were \log_{10} - or square root-transformed prior to imputation (24). Ten multiple imputed data sets were generated, and pooled estimates were reported. Owing to their positively skewed distribution, HDL cholesterol, triglycerides, hs-CRP, and waist circumference were logarithmically transformed (\log_{10}), while HOMA-IR was square root-transformed.

Separate multiple linear regression models were performed to test the independent association of MVPA, CPM, and ST with cardiometabolic health markers (HOMA-IR, fasting insulin, fasting plasma glucose, 2-h glucose, HbA_{1c} , waist circumference, triglycerides, total cholesterol, HDL cholesterol, and LDL cholesterol) and CRP with adjustment for potential confounders.

Model 1 was adjusted for age, sex, ethnicity (Caucasian, Asian, black, Arabic, Hispanic, or other), smoking (daily, less than weekly, and no smoking), accelerometer wear time, intervention center, sleep duration, body fat%, education level (no formal education, primary/junior school, secondary school, secondary vocational education, higher vocational education, or university education), and household income (less than £9,360, £9,360–12,479.99, £12,480–15,599.99, £15,600–18,719.99, £18,720–22,879.99, £22,880–27,559.99, £27,560–32,759.99, £32,760–41,079.99, £41,080–53,559.99, and £53,560 or more per year). Model 2 was additionally adjusted for ST when MVPA was the main exposure variable or for MVPA when ST was the main exposure variable.

Sex, age (<45.9, 46–54.9, and >55 years), BMI, and intervention center differences in the associations between MVPA, ST, or CPM and each cardiometabolic risk factor were tested for by adding interaction terms to the model.

A variance inflation factor of <4 confirmed that multicollinearity was not a concern (25). Square root and log transformations were directly compared across cardiometabolic markers and results of linear regression analysis presented as standardized β -coefficients. Data were analyzed using the Statistical Package for the Social Sciences, version 22.0 (SPSS, Chicago, IL), and α was set to $P < 0.05$.

RESULTS

Table 1 displays the descriptive physical and biochemical characteristics of the 2,326 participants (32% male, age 52.2 ± 11.5 years). Waking accelerometer wear time was 928.40 ± 83.37 and 933.10 ± 83.85 min/day for men and women, respectively. Mean ST was 617.54 ± 98.06 for men and 579.46 ± 91.76 min/day for women. Mean MVPA was 31.58 ± 20.62 min/day for men and 26.18 ± 17.03 min/day for women. Only 50% of participants met the recommended guidelines of 30 min of MVPA per day (26).

Tables 2–4 show the standardized regression coefficients of CPM, MVPA, and ST with the cardiometabolic risk factors.

CPM

Table 2 shows the standardized regression coefficients of CPM with the cardiometabolic risk factors. After adjustment

for age, sex, ethnicity, smoking, accelerometer wear time, intervention center, sleep duration, body fat%, education level, and household income, CPM had significant inverse associations with HOMA-IR, waist circumference, fasting insulin, 2-h glucose, triglycerides, and CRP and a positive association with HDL cholesterol.

MVPA

Table 3 shows the standardized regression coefficients for MVPA with the cardiometabolic risk factors. In model 1, adjusted for age, sex, ethnicity, smoking, accelerometer wear time, intervention center, sleep duration, body fat%, education level, and household income, MVPA was significantly and negatively associated with HOMA-IR, waist circumference, fasting insulin, 2-h glucose, triglycerides, and CRP and positively associated with HDL cholesterol.

After adjustment for ST in model 2, the association with HDL cholesterol was lost. Associations with other biochemical factors were slightly attenuated, but all remained significant.

Sedentary Time

Table 4 shows the standardized regression coefficients for ST with the cardiometabolic risk factors. In model 1, adjusted for age, sex, ethnicity, smoking, accelerometer wear time, intervention center, sleep duration, body fat%, education level, and household income, ST was positively associated with waist circumference, systolic BP, diastolic BP, mean arterial pressure, fasting insulin, 2-h glucose, HOMA-IR, triglycerides, and CRP and negatively associated with HDL. After adjustment for MVPA in model 2, associations with 2-h plasma glucose were no longer significant.

Fasting plasma glucose, total cholesterol, and LDL cholesterol were not associated with any of our exposure variables.

Two-way interactions indicate that associations between MVPA and fasting insulin were greater in the older age-group (Table 3). We did not observe any significant sex, center, or BMI interactions between measures of PA or ST and any cardiometabolic risk factors.

CONCLUSIONS

This study investigated the associations between objectively measured PA and ST and metabolic variables in a worldwide sample of overweight and obese adults

Table 1—Descriptive, metabolic, and PA characteristics of 2,326 adults with prediabetes from PREVIEW

Characteristics	Female (n = 1,570)	Male (n = 755)
Age (years)	51.6 ± 11.6	53.5 ± 11.6
Height (m)	1.64 ± 0.07	1.77 ± 0.07
Weight (kg)	95.88 ± 20.33	108.87 ± 20.98
BMI (kg · m ⁻²)	35.70 ± 6.76	34.55 ± 6.01
Fat (%)	46.45 ± 5.85	36.67 ± 6.92
Waist (cm)	107.7 ± 14.6	116.8 ± 14.5
Systolic BP (mmHg)	126.98 ± 5.88	133.10 ± 14.8
Diastolic BP (mmHg)	77.07 ± 11.15	80.88 ± 9.96
Fasting insulin (mU · L ⁻¹)	12.53 ± 6.54	14.03 ± 6.68
Fasting plasma glucose (mmol · L ⁻¹)	6.08 ± 0.67	6.33 ± 0.66
2-h plasma glucose (mmol · L ⁻¹)	7.64 ± 2.21	7.73 ± 2.24
HbA _{1c} (mmol · L ⁻¹)	36.61 ± 4.03	36.73 ± 4.06
HbA _{1c} (%)	5.50 ± 0.37	5.51 ± 0.37
HOMA-IR	3.44 ± 1.90	4.00 ± 2.01
Triglycerides (mmol · L ⁻¹)	1.45 ± 0.77	1.62 ± 0.82
Total cholesterol (mmol · L ⁻¹)	5.25 ± 0.99	5.03 ± 0.97
HDL cholesterol (mmol · L ⁻¹)	1.33 ± 0.29	1.15 ± 0.23
LDL cholesterol (mmol · L ⁻¹)	3.28 ± 0.84	3.16 ± 0.85
CRP (mg · L ⁻¹)	4.81 ± 4.02	3.46 ± 3.35
Accelerometer variables		
Waking wear time (min · day ⁻¹)	933.10 ± 83.85	928.40 ± 83.37
Sleep (min · day ⁻¹)	474.69 ± 80.26	471.83 ± 85.16
ST (min · day ⁻¹)	579.46 ± 91.76	617.54 ± 98.06
Light activity (min · day ⁻¹)	320.84 ± 82.62	280.12 ± 78.20
Moderate activity (min · day ⁻¹)	25.30 ± 6.88	30.33 ± 20.33
MVPA (min · day ⁻¹)	26.18 ± 17.03	31.58 ± 20.62
CPM	294.30 ± 96.77	297.98 ± 17.28
Ethnicity (%)		
Caucasian	86.02	89.9
Asian	2.66	2.71
Black	1.82	1.2
Arabic	0.2	0.3
Hispanic	2.33	1.38
Other	6.61	4.25
Smoking (%)		
Yes (daily)	10.72	9.57
Sometimes (less than weekly)	3.3	2.93
No	85.98	87.48
Education (%)		
No formal education	0.5	—
Primary/junior school	2.1	2.8
Secondary school	15.1	14.2
Secondary vocational education	17.7	19.2
Higher vocational education	16.4	18.6
University education	40.5	35.6
Other	7.6	9.7
Household income per year (%)		
Less than £9,360	8.8	5.3
£9,360–12,479.99	5.8	2.9
£12,480–15,599.99	5.1	2.5
£15,600–18,719.99	6.5	3.5
£18,720–22,879.99	5.7	4.6
£22,880–27,559.99	8.1	5.9
£27,560–32,759.99	10.6	8.1
£32,760–41,079.99	11.2	10.9
£41,080–53,559.99	13.4	14.2
£53,560 or more	24.9	42

Data are means ± SD unless otherwise stated.

(BMI >25 kg/m²) with prediabetes confirmed through an OGTT (27). To our knowledge, this is the first international investigation of associations between objectively measured PA and ST with cardiometabolic risk factors in a population that exclusively meet the criteria for prediabetes (impaired fasting glucose [5.6–6.9 mmol · L⁻¹] or impaired glucose tolerance [7.8–11 mmol · L⁻¹]). Previous studies conducted in high-risk populations have included participants based on a risk score questionnaire (5) or family history of type 2 diabetes (9,13), or participants with diabetes and prediabetes have been combined (14). Consequently, physical characteristics and metabolic parameters reported in these studies varied substantially from those in the current study. In our population of participants with prediabetes, MVPA was negatively associated with HOMA-IR, waist circumference, fasting insulin, 2-h glucose, triglycerides, and CRP after accounting for potential confounders (sleep duration and ST). Our study also demonstrated that total PA volume (CPM) was at least as strongly associated with the aforementioned risk factors as was MVPA.

Before ST was controlled for in model 2, total PA volume also accounted for greater variance in cardiometabolic risk factors than did MVPA. Hence, it appears that in this population, although both are significant, the accumulation of total PA over a day is a stronger indicator of insulin resistance and some related cardiometabolic risk factors than MVPA.

Previous studies in populations with a family history of diabetes and newly diagnosed diabetes have shown that total energy expenditure spent on PA (14), CPM (13), and MVPA (9) were negatively associated with waist circumference, fasting serum triglycerides, systolic BP, fasting plasma glucose, fasting plasma insulin, HOMA-IR, and a clustered metabolic risk score. In agreement with our findings, Ekelund, Griffin, and Wareham (13) reported that total counts · day⁻¹ was more strongly associated with clustered risk and individual cardiometabolic risk factors than was MVPA. In the general population, Balkau et al. (28) reported associations between MVPA and total PA with insulin resistance using the gold standard clamp technique for determining insulin sensitivity. However, after adjustment for total PA, associations with MVPA were lost (28). These findings, in

Table 2—Standardized β -coefficients for associations between total PA (CPM) and cardiometabolic risk factors

Characteristics	β	95% CI	P	R ²
Waist (cm)	−0.179	−0.224, −0.134	*	0.2057
Systolic BP (mmHg)	−0.033	−0.074, 0.007		0.1927
Diastolic BP (mmHg)	−0.02	−0.057, 0.017		0.3448
Fasting insulin (mU · L ^{−1})	−0.139	−0.183, −0.096	*	0.1293
Fasting glucose (mmol · L ^{−1})	−0.028	−0.074, 0.018		0.1647
2-h plasma glucose (mmol · L ^{−1})	−0.088	−0.131, −0.045	*	0.1139
HbA _{1c} (%)	−0.049	−0.098, −0.001		0.1337
HOMA-IR	−0.151	−0.194, −0.107	*	0.1306
Triglycerides (mmol · L ^{−1})	−0.117	−0.162, −0.071	*	0.0812
Total cholesterol (mmol · L ^{−1})	−0.007	−0.053, 0.038		0.095
HDL (mmol · L ^{−1})	0.088	0.048, 0.129	*	0.1856
LDL (mmol · L ^{−1})	0.002	−0.045, 0.049		0.0566
CRP (mg · L ^{−1})	−0.104	−0.146, −0.062	*	0.2657

Adjusted for age, sex, ethnicity, smoking, household income, education level, body fat%, wear time, sleep time, and intervention center. * $P < 0.001$.

keeping with the current study, would support the hypothesis that the accumulation of total PA volume accounts for greater variance in insulin resistance and

some related cardiometabolic risk factors than does MVPA.

The current study also demonstrated that after confounders were controlled

Table 3—Standardized β -coefficients for associations between MVPA (min · day^{−1}) and cardiometabolic risk factors

Characteristics	β	95% CI	P	R ²
Model 1				
Waist (cm)	−0.177	−0.122, −0.134	***	0.2046
Systolic BP (mmHg)	−0.005	−0.047, 0.083		0.1915
Diastolic BP (mmHg)	−0.007	−0.044, 0.031		0.3441
Fasting insulin (mU · L ^{−1})	−0.115	−0.158, −0.072	***	0.1237
Fasting glucose (mmol · L ^{−1})	−0.028	−0.072, 0.017		0.1645
2-h plasma glucose (mmol · L ^{−1})	−0.069	−0.112, −0.025	**	0.1108
HbA _{1c} (%)	−0.046	−0.096, 0.004		0.1334
HOMA-IR	−0.122	−0.166, −0.078	***	0.1235
Triglycerides (mmol · L ^{−1})	−0.091	−0.138, −0.044	***	0.0762
Total cholesterol (mmol · L ^{−1})	−0.01	−0.056, 0.035		0.0952
HDL (mmol · L ^{−1})	0.055	0.009, 0.101	*	0.1808
LDL (mmol · L ^{−1})	0.002	−0.044, 0.048		0.0567
CRP (mg · L ^{−1})	−0.086	−0.127, −0.045	***	0.262
Model 2				
Waist (cm)	−0.127	−0.173, −0.081	***	0.2215
Systolic BP (mmHg)	0.02	−0.026, 0.067		0.1961
Diastolic BP (mmHg)	0.011	−0.03, 0.053		0.3467
Fasting insulin (mU · L ^{−1}) main effect	−0.078	−0.127, −0.03	**	0.1337
25–45.9 years (reference)	—	—	—	—
46.0–54.9 years	0.025	−0.120, 0.170		0.1313
55.0–71 years	−0.098	−0.195, −0.001	*	0.1313
Fasting glucose (mmol · L ^{−1})	−0.014	−0.063, 0.035		0.1659
2-h plasma glucose (mmol · L ^{−1})	−0.053	−0.1, −0.006	*	0.1129
HbA _{1c} (%)	−0.036	−0.095, 0.023		0.1345
HOMA-IR	−0.08	−0.129, −0.031	**	0.1368
Triglycerides (mmol · L ^{−1})	−0.067	−0.117, −0.017	**	0.0804
Cholesterol (mmol · L ^{−1})	−0.013	−0.064, 0.037		0.0953
HDL (mmol · L ^{−1})	0.028	−0.081, 0.383		0.1867
LDL (mmol · L ^{−1})	0	−0.053, −0.052		0.0572
CRP (mmol · L ^{−1})	−0.061	−0.108, −0.015	**	0.2665

Model 1 adjusted for age, sex, ethnicity, smoking, household income, education level, body fat%, wear time, sleep time, and intervention center. Model 2 additionally adjusted for ST. * $P < 0.05$; ** $P < 0.01$; *** $P < 0.001$.

for, ST was positively associated with waist circumference, systolic BP, diastolic BP, fasting insulin, HOMA-IR, triglycerides, CRP, and HDL cholesterol independent of time spent in MVPA.

In agreement with our findings, Henson et al. (5) reported positive associations between ST, 2-h plasma glucose, triglycerides, and HDL cholesterol, independent of total PA, in a population at risk for type 2 diabetes (5). Similarly, positive associations between ST, waist circumference, insulin, HOMA-IR, and HDL cholesterol were reported in an older group with newly diagnosed type 2 diabetes (22). There is a lack of consensus on the role of ST in pathology of cardiometabolic risk factors. Some authors have reported that once total energy expenditure is taken into account, the associations between ST and cardiometabolic health are lost. These findings suggest that ST reduces total energy expenditure by displacing other activities that are more energy costly (28,29). Other investigators have found that the mechanism linking ST to glucose metabolism and metabolic health differ from those of PA and may be related to static posture and unloading of large skeletal muscle groups when seated (7,30). ST is thought to affect glucose homeostasis and lipid metabolism by reducing muscle GLUT4 content and insulin-stimulated glucose uptake (6) while also reducing lipoprotein lipase activity, leading to impaired triglyceride and HDL cholesterol metabolism (7).

Further, light activity is associated with marked improvements in cardiometabolic health (8), while interventions have shown that replacing ST with postural changes such as standing or light ambulatory activity can improve glycemic control to a greater extent than structured exercise of the same energy cost (31). In the present analysis, ST was associated with HDL cholesterol; however, associations between MVPA and HDL cholesterol were lost after controlling for ST. Similarly, ST was positively associated with systolic and diastolic BP independent of MVPA, while no association was found between CPM or MVPA with either BP variable. Associations of PA and ST with BP variables reported by observational studies have been inconsistent (8,12,14). However, experimental studies in overweight/obese populations have suggested that a reduction of ST and the interruption of prolonged sedentary

Table 4—Standardized β -coefficients for associations between ST (min \cdot day⁻¹) and cardiometabolic risk factors

Characteristics	β	95% CI	P	R ²
Model 1				
Waist (cm)	0.215	0.146, 0.267	***	0.2088
Systolic BP (mmHg)	0.078	0.026, 0.131	**	0.1957
Diastolic BP (mmHg)	0.057	0.007, 0.106	*	0.3464
Fasting insulin (mU \cdot L ⁻¹)	0.155	0.092, 0.219	***	0.1291
Fasting glucose (mmol \cdot L ⁻¹)	0.052	-0.004, 0.108		0.1656
2-h plasma glucose (mmol \cdot L ⁻¹)	0.072	0.015, 0.129	*	0.1106
HbA _{1c} (%)	0.047	-0.022, 0.117		0.1332
HOMA-IR	0.175	0.114, 0.236	***	0.1316
Triglycerides (mmol \cdot L ⁻¹)	0.106	0.052, 0.16	***	0.0766
Total cholesterol (mmol \cdot L ⁻¹)	-0.006	-0.062, 0.051		0.0953
HDL (mmol \cdot L ⁻¹)	-0.103	-0.165, -0.042	***	0.1859
LDL (mmol \cdot L ⁻¹)	-0.007	-0.062, 0.048		0.0568
CRP (mg \cdot L ⁻¹)	0.106	0.039, 0.172	**	0.2632
Model 2				
Waist (cm)	0.165	0.109, 0.221	***	0.2215
Systolic BP (mmHg)	0.086	0.028, 0.143	**	0.1961
Diastolic BP (mmHg)	0.061	0.006, 0.116	*	0.3467
Fasting insulin (mU \cdot L ⁻¹)	0.126	0.055, 0.198	**	0.1337
Fasting glucose (mmol \cdot L ⁻¹)	0.047	-0.014, 0.108		0.1659
2-h plasma glucose (mmol \cdot L ⁻¹)	0.053	-0.009, 0.114		0.1129
HbA _{1c} (%)	0.034	-0.045, 0.113		0.1345
HOMA-IR	0.145	0.077, 0.213	***	0.1368
Triglycerides (mmol \cdot L ⁻¹)	0.08	0.023, 0.138	**	0.0804
Total cholesterol (mmol \cdot L ⁻¹)	-0.011	-0.074, 0.052		0.0953
HDL (mmol \cdot L ⁻¹)	-0.093	-0.163, -0.022	*	0.1867
LDL (mmol \cdot L ⁻¹)	-0.007	-0.069, 0.055		0.0572
CRP (mg \cdot L ⁻¹)	0.083	0.01, 0.156	*	0.2665

Model 1 adjusted for age, sex, ethnicity, smoking, household income, education level, body fat%, wear time, sleep time, and intervention center. Model 2 additionally adjusted for MVPA. * $P < 0.05$; ** $P < 0.01$; *** $P < 0.001$.

bouts with light or moderate activity are associated with improved systolic and diastolic BP (32).

While PA guidelines continue to focus on participation in MVPA (26), our data demonstrated that in a population with prediabetes, total volume of PA was as strongly associated with cardiometabolic health as was MVPA. The implications of this finding may be important considering the levels of PA in this population. Light activity may be more readily adopted by individuals with prediabetes, particularly if they are physically inactive, overweight/obese, or reluctant to engage in structured exercise (31).

The lack of associations observed for all exposure variables with fasting glucose is consistent with previous research (5,8). This finding reflects the fact that PA predominantly affects peripheral insulin sensitivity, which is responsible for lowering blood glucose levels after an OGTT, when most of the glucose is taken up by skeletal muscle. Conversely, in a fasted state, peripheral insulin sensitivity has little effect on plasma glucose (33).

The strengths of this study are the implementation of a 24-h accelerometer wear-time protocol that resulted in a greater mean waking wear time (15.5 h) than seen in many studies of a similar nature (5,18,22). Longer monitoring duration provides greater reliability of average activity estimates. This approach also allowed an objective assessment of sleep time using an algorithm to detect sleep onset and wake time from a 24-h waist-worn accelerometer trace (19). This allowed the study team to control for the confounding effects of sleep on cardiometabolic risk factors (34).

Type 2 diabetes is high on public health agendas, with attention on the prevention or delay of diabetes onset and the management of cardiometabolic risk factors (35,36). National and international guidelines focus on, first, identifying high-risk individuals and, second, controlling modifiable risk factors such as body weight, diet, ST, and PA through targeted interventions (37,38). This study provides new evidence of associations of PA and ST with cardiometabolic markers in a

population to whom the results are most applicable.

While this study has numerous strengths, it is also important to acknowledge its limitations. First, the cross-sectional design does not allow insight into the direction of causality between each exposure variable and markers of cardiometabolic health, and although we controlled for many potential confounding factors, we did not account for dietary intake or alcohol consumption, which may have influenced our results. Second, although accelerometers offer more robust assessments of PA than self-report (39), they are not without limitation. Hip-worn accelerometers capture most movement during locomotion but cannot account for upper-body movement, movement that occurs during activities such as cycling or weight lifting (40), or distinguish between light-intensity activities such as sitting and standing. Furthermore, the accelerometer is removed during water-based activities and contact sports. Therefore, PA may be underestimated. Given the fixed nature of accelerometer-derived variables (sleep, light activity, ST, and MVPA) as proportions of 24 hours, time spent in behaviors within the day are inherently collinear; every increase in time spent in one behavior unavoidably causes a decrease in the time spent in one or a combination of the other behaviors. Thus, it is not possible to include all subcomponents of the day (sleep, ST, light activity, and MVPA) in a regression model without violating collinearity assumptions. Consequently, in the current study it is not possible to say with certainty that the positive associations observed between cardiometabolic risk factors and ST are truly independent and not, in fact, negative associations with light activity. Finally, participants in this study were volunteers for a lifestyle intervention from the eight study sites worldwide, approximately 50% of whom were between 55 and 70 years of age. While this may limit the applicability of the current findings to this older age range, it is this group who is at greater risk of type 2 diabetes and is perhaps more likely to participate in such interventions given their greater availability of time compared with those still employed.

In conclusion, this study provides new evidence that in a large diverse population of adults with prediabetes, objective measures of PA and ST are associated with

markers of cardiometabolic health. Furthermore, associations with total PA volume are at least as strong as with MVPA. Taken together, these findings suggest that replacing ST with light activity may provide a practical approach to improve cardiometabolic health in a population with low engagement in MVPA.

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Duality of Interest. I.M. is an International Life Sciences Institute Europe member of the Dietary Carbohydrates Task Force and a member of the expert group on "Efficacy Markers of Diabetes Risk" and expert group on "Carbohydrate-Based Recommendations as a Basis for Dietary Guidelines: A Scientific Review." In all cases, travel and subsistence are paid but not attendance fees. I.M. received honorarium as part of the Nature Publishing Group (Springer Nature) and as Editor of the *International Journal of Obesity*. I.M. has had research travel and accommodation reimbursed for a Nestle Research Centre Consultancy for Nutrition in the Life Cycle (honorarium paid to the University of Nottingham). I.M. received honorarium from Mars Incorporated—Waltham Centre for Pet Nutrition for providing peer-review of pet nutrition research (amount received per annum less than £5,000). I.M. has had travel and subsistence costs reimbursed by Mars UK/Europe as a Member of Nutrition Advisory Board and Health and Wellbeing Committee (honorarium paid to the University of Nottingham). I.M. has had travel and subsistence costs reimbursed as an Ikea Member of the Science and Health Committee (honorarium paid to the University of Nottingham). S.D.P. holds the Fonterra Chair in Human Nutrition at the University of Auckland. J.B.-M. is the President of the Glycemic Index Foundation, Director of the Sydney University Glycemic Index Research Service, and the author of books about the glycemic index of foods. No other potential conflicts of interest relevant to this article were reported.

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