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Job stress, especially high job strain and effort-reward imbalance, was modestly associated with increased risk of physical inactivity, even after controlling for individual time-invariant attributes. Low job control and low reward amplified the association of physical inactivity with high job demands and high effort, respectively. Policy support is needed to support physical activity among workers facing job stress.

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Key terms: [cohort study](#); [cohort survey](#); [effort-reward imbalance](#); [ERI](#); [fixed-effects model](#); [Japan](#); [job strain](#); [job stress](#); [leisure-time physical inactivity](#); [physical inactivity](#)

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The association between job stress and leisure-time physical inactivity adjusted for individual attributes: evidence from a Japanese occupational cohort survey

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Objective We examined the association between job stress and leisure-time physical inactivity, adjusting for individual time-invariant attributes.

Methods We used data from a Japanese occupational cohort survey, which included 31 025 observations of 9871 individuals. Focusing on the evolution of job stress and leisure-time physical inactivity within the same individual over time, we employed fixed-effects logistic models to examine the association between job stress and leisure-time physical inactivity. We compared the results with those in pooled cross-sectional models and fixed-effects ordered logistic models.

Results Fixed-effects models showed that the odds ratio (OR) of physical inactivity were 22% higher for those with high strain jobs [high demands/low control; OR 1.22, 95% confidence interval (95% CI) 1.03–1.43] and 17% higher for those with active jobs (high demands/high control; OR 1.17, 95% CI 1.02–1.34) than those with low strain jobs (low demands/high control). The models also showed that the odds of physical inactivity were 28% higher for those with high effort/low reward jobs (OR 1.28, 95% CI 1.10–1.50) and 24% higher for those with high effort/high reward jobs (OR 1.24, 95% CI 1.07–1.43) than those with low effort/high reward jobs. Fixed-effects ordered logistic models led to similar results.

Conclusion Job stress, especially high job strain and effort–reward imbalance, was modestly associated with higher risks of physical inactivity, even after controlling for individual time-invariant attributes.

Key terms cohort study; effort–reward imbalance; ERI; fixed-effects model; Japan; job strain.

It is widely known that leisure-time physical inactivity is a major risk factor for chronic disease, morbidity, and other health outcomes. However, the reduction in a sedentary lifestyle has been modest in advanced countries (1–4). Meanwhile, studies have suggested that exposure to job stress can be a potential contributor to leisure-time physical inactivity, as well as to other health-risk behaviors (5–12). It can be hypothesized that job stress may result in fatigue as well as general passivity and apathy, which may spill over to leisure time and impede the implementation of exercise intentions, increasing the likelihood of physical inactivity in leisure time (6,

13, 14). Hence, a further understanding of the association between job stress and physical inactivity can help design policy strategies to enhance workers' health.

The job demands–control (JDC) model focuses on the interaction of job demands and control as a key determinant of workers' health outcomes (15). Based on this model, it is reasonable to predict that high strain jobs, which are characterized by high demands and low control, would result in increased risks of physical inactivity. The effort–reward imbalance (ERI) model may also have implications for physical inactivity since it argues that an imbalance between higher effort spent on

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work and lower reward obtained from it has a stressful impact on workers (16). Based on the JDC and/or ERI models, an increasing number of studies have examined the association between job stress and physical inactivity (5–12, 17–19). However, the results have been generally inconclusive; some studies have provided evidence supportive of a positive, albeit modest, association between job stress and physical inactivity (5–12), while others have been skeptical of such an association (17–20).

One potential reason for these mixed results may be that most preceding studies have relied on a cross-sectional analysis of the data (5, 8, 10, 11, 17–20). Job stress is usually measured based on the respondent's self-reported evaluations of workplace characteristics, which are most likely affected by individual attributes in addition to objective workplace characteristics. Physical inactivity is also likely affected by individual attributes; for instance, several studies have identified some aspects of personality traits as physical activity correlates (21).

A key limitation of a cross-sectional analysis is that its estimation results cannot be free from biases related to unobserved time-invariant confounders. Although cross-sectional analysis can control for observed time-invariant confounders (such as gender and educational attainment), it cannot control for unobserved ones (such as personality traits and other inherent individual characteristics, chronic disease, and experience in childhood). This is most likely to lead to biased estimation results. For example, if individuals with a certain unobserved attribute (such as high neuroticism) tend to feel more job stress than others, the observed association between job stress and physical inactivity in a cross-sectional analysis may reflect the association of physical inactivity with that individual attribute, rather than job stress per se.

A conventional method to overcome these problems is to employ the fixed-effects method for analysis, which focuses on within-individual variations (22–25). This method enables us to obtain unbiased estimators by controlling for individual time-invariant attributes – that is, fixed effects – both observed and unobserved. A key focus is on whether the association, which is observed in a cross-sectional analysis, remains significant even after controlling for time-invariant attributes in a fixed-effects analysis. An increasing number of studies have been employing the fixed-effects regression method to examine the determinants of mental health and other health variables (23–25). To the best of our knowledge, however, there have been few attempts to apply this method to examine the association between job stress and physical inactivity, with the notable exception of Kouvonen et al's work (12).

In the present study, we focused on the evolution of job stress and leisure-time physical inactivity within the same individual over time, using longitudinal data obtained from an occupational Japanese cohort survey.

Explicitly, we employed fixed-effects logistic models to examine the association between job stress and leisure-time physical inactivity, controlling for individual time-invariant attributes. Based on Kouvonen et al's work (12), we considered both the JDC and ERI models. We further investigated how each demands–control and effort–reward combination was associated with physical inactivity. Some previous studies have compared the associations with physical inactivity across job demands–control combinations (6, 8, 9, 11, 17, 18, 20), but effort–reward combinations have rarely been analyzed. This method allowed us to assess the relative importance of each job stress aspect as physical inactivity correlates. Moreover, our dataset, which contained data from 2–4 waves, allowed us to obtain more reliable information about the association between job stress and physical inactivity. This was as compared to previous cohort studies, most of which compared data at the baseline and one follow-up time only (6, 9).

Methods

Study sample

We used panel data from four survey waves of an occupational cohort study on social class and health in Japan (Japanese Study of Health, Occupation, and Psychosocial Factors Related Equity or J-HOPE). The first wave was conducted from October 2010 to December 2011; following waves were conducted approximately one year after the previous one. Data were collected from annual worksite health check-ups, which are mandatory for all Japanese employees. The recruitment periods varied among the study sites; the health check-ups were conducted in a fixed month every year, for all employees, in each employee's birth month.

The study population consisted of employees working for 13 firms, 3 of which participated only in the first three waves. The surveyed firms covered 12 industries, while the surveyed respondents were classified into nine occupation types. The original sample consisted of 10 753, 11 405, 10 977, and 6553 respondents in the first to the fourth waves, respectively (response rates: 77.0%, 81.7%, 78.6%, and 67.0%, respectively). The original dataset included 39 683 observations of 14 140 individuals (10 550 men and 3590 women), who joined at least one wave. The attrition rates were 18.3%, 13.2%, and 16.5% in the second, third, and fourth waves, respectively. The respondents were aged 18–76 years [mean 41.5 (SD 10.3) years].

To compare cross-sectional and fixed-effects models, we focused on individuals who joined at least two consecutive waves. Excluding respondents with missing

job stress indicators and other important variables, we utilized 31 025 observations of 9871 individuals (7593 men and 2285 women; 78.2% of the original sample observations; 69.8% of the original sample individuals). Among these, 3382 joined all four waves, and 5020 and 1469 joined only three and two waves, respectively. The present analysis was conducted with the J-HOPE dataset as of 22 August 2014.

Measures

Physical inactivity. J-HOPE asked the respondents to report their leisure-time physical activity on a 4-point scale: (i) none; (ii) low (ie, mild exercise without breathlessness or heart palpitations) ≥ 1 /week; (iii) intense (ie, heavy exercise with breathlessness, heart palpitation, or sweating for ≥ 20 minutes) 1–2/week; and (iv) intense physical activity ≥ 3 /week. Based on the definition of physical activity in previous studies (eg, 1, 2, 6, 8) we constructed a binary variable of physical inactivity by allocating one to the answer “none” and zero to the others. Alternatively, we constructed a 4-point-score categorical variable of physical inactivity measured on a 4-point scale by reversing the order (4 = none to 1 = intense physical activity ≥ 3 /week), rather than setting up a single cut off score for physical inactivity. These self-reported responses could not be directly converted into metabolic equivalent tasks (MET). Hence, we estimated regression models to explain each of 11 health-risk factors (see the section on health-risk factors) by the above-defined binary variable of physical inactivity, to assess its relevance to workers’ health.

Job demands and control. We utilized the items investigating job demands and control from the Japanese version of the Job Content Questionnaire (JCQ). The JCQ is based on the JDC model, and includes scales related to job demands (five items) and job control (nine items) rated on a 4-point scale (1=strongly disagree to 4=strongly agree) (25). The internal consistency, reliability, and validity of the Japanese version of the JCQ are acceptable (26). In the present sample, we summed the responses to these items into single indices of job demands (range: 12–48) and control (range: 24–96), based on the standardized formulae (26, 27). We then used the study-specific median scores as cut-off points for high and low demands, and high and low job control. Finally, we categorized four job types: 1) low demands and high control (low strain); 2) high demands and high control (active); 3) low demands and low control (passive); and 4) high demands and low control (high strain) (28).

Extrinsic effort and reward. To assess effort and reward, we utilized the data collected from a simplified Japanese version of the Effort–Reward Imbalance Questionnaire

(ERIQ). The ERIQ was developed based on the ERI model (29). Its Japanese version and that of the simplified ERIQ (30) used in the present study have acceptable internal consistency, reliability, and validity scores (31, 32). The simplified version includes sub-scales for extrinsic effort (three items) and reward (seven items) rated on a 4-point scale (1=strongly disagree to 4=strongly agree). We summed the responses into single indices for effort (range: 3–12) and reward (range: 7–28), based on the standardized formulae (30–32). We then used the study-specific median scores as cut-off points for high and low effort and high and low reward. Finally, we categorized four job types: (i) low effort/high reward, (ii) high effort/high reward, (iii) low effort/low reward, and (iv) high effort/low reward.

Health-risk factors. We considered 11 health-risk factors including psychological distress. The data were collected from annual worksite health check-ups; the K6 score (33) was an indicator of psychological distress, constructed from the survey responses. The ten health-risk factors included systolic blood pressure, diastolic blood pressure, triglyceride level, total cholesterol level, high-density lipoprotein (HDL) cholesterol level, low-density lipoprotein (LDL) cholesterol level, fasting blood sugar level, hemoglobin A1c (HbA1c) level, body mass index (BMI), and waist circumference.

To construct the K6 score, we first obtained the respondents’ assessments of psychological distress using a 6-item psychological distress questionnaire: “During the past 30 days, about how often did you feel a) nervous, b) hopeless, c) restless or fidgety, d) so depressed that nothing could cheer you up, e) that everything needed added effort, and f) was worthless?”; responses were rated on a 5-point scale (0=none of the time to 4=all of the time). We then calculated the sum of the reported scores (range: 0–24) and defined it as the K6 score. Higher K6 scores reflect higher levels of psychological distress. Its reliability and validity have been demonstrated on a Japanese sample (34, 35).

Covariates

We used both individual time-invariant and -variant variables as covariates. For individual time-invariant variables, we used gender, educational attainment (high school or below, junior college, college, and graduate school), and 13 firm indicators. Among individual time-variant variables, the most important one was hours worked per week. As suggested by preceding studies (36, 37), longer hours worked are expected to restrain physical inactivity through reduced leisure time, regardless of workplace stress. The survey asked the respondents to choose from five groups related to hours worked per week (≤ 30 , 31–40, 41–50, 51–60, and ≥ 61

hours). For simplicity, we assumed 20, 35, 45, 55, and 65 hours for each bracket and used them as scores of a continuous variable of hours worked. In addition, we considered nine job types (such as manager, clerk, and factory worker), six income brackets (≤ 2.99 , 3–4.99, 5–7.99, 8–9.99, 10–14.99, and ≥ 15 million yen), and age (<30 , 30s, 40s, 50s, and ≥ 60 years or above) as individual time-variant variables. We also included indicator variables of the four waves to control for wave-specific factors.

Statistical analysis

After descriptive analysis, which aimed at summarizing the basic features of each aspect of job stress, we conducted three analyses. First, we assessed the relevance of physical inactivity in workers' health, by estimating linear regression models to explain the 11 health-risk factor variables by physical inactivity along with covariates. This analysis was needed because our physical inactivity measure was based on the respondent's self-assessments and its relevance in workers' health was not necessarily warranted. Second, as a main analysis, we examined the association between job stress and physical inactivity. Specifically, we estimated physical inactivity by three job types – high demands/high control (active), low demands/low control (passive), and high demands/low control (high strain) – considering low demands/high control (low strain) as a reference, based on the JDC model. We also estimated an ERI version of this model, with high effort/high reward, low effort/low reward, and high effort/low reward as regressors, and low effort/high reward as a reference. In these models, we used a binary variable of physical inactivity and employed logistic regression models. Finally, we examined the robustness of the results of these logistic models. To this end, we replaced a binary variable of physical inactivity with its 4-point categorical variable described earlier and estimated ordered logistic models to explain physical inactivity by the same regressors used in the logistic models.

In all of these regression analyses, we employed fixed-effects models in order to control for individual time-invariant attributes, whether observed or unobserved. To enable explanation of each health-risk variable, which was a continuous variable, using the above models, we regressed each mean-centered health-risk variable on all mean-centered regressors (22). For the logistic fixed-effects models in the second analysis, we used the maximum likelihood estimation conditional on the sum of events (ie, the amount of physical inactivity during the waves), which provided unbiased estimators (38). For the ordered logistic fixed-effects models, we employed a recently developed method (39), which provided unbiased estimators as in the case of logistic fixed-effects models. For both logistic and ordered

logistic models, respondents who stayed in the same physical activity status were automatically dropped from the analysis, because these models concentrated on within-individual variations. For all of these fixed-effects models, we compared their results with those in pooled cross-sectional models to assess the importance of controlling for individual time-invariant attributes.

Results

Table 1 summarizes basic characteristics of respondents at baseline, showing that 3741 (35.1%) of the respondents were physically inactive in leisure time. Women were less active than men. Educational attainment and income were positively associated with physical activity, while smoking was negatively associated with it. We controlled for these factors in the regression analysis.

The key features of the four job stress variables are presented in table 2 in terms of (i) the ranges, means, and standard deviations of their scores, (ii) pairwise correlations with each other, and (iii) prevalence of physical inactivity by their high and low levels. This table shows high correlations between job demands and effort (0.617) and between job control and reward (0.345), suggesting that these pairs of concepts – as well as the two models of job stress (JDC and ERI models) – potentially overlap with each other, respectively. This table also shows that the prevalence of physical inactivity was positively associated with high job demands, low job control, high effort, and low reward, suggesting that each aspect of job stress was independently associated with physical inactivity. In addition, the prevalence of physical inactivity was very close between high job demands and high effort, and

Table 1. Basic characteristics of the respondents at baseline.^a

	All (N=9871)			Physically active (N=3741)			Physically inactive (N=6130)		
	%	Mean	SD	%	Mean	SD	%	Mean	SD
Men	76.9			83.4			72.9		
Women	23.1			16.6			27.1		
Education									
High school or less	39.2			36.4			41.0		
Junior college	16.3			13.6			18.0		
College	33.8			37.9			31.3		
Graduate school	10.6			12.2			9.6		
Smoking	28.4			25.7			30.1		
Age (years)	40.3	10.4		41.0	10.6		39.9	10.3	
Income (¥ million/year)	6.9	3.5		7.3	3.5		6.7	3.3	
Hours worked per week	41.9	11.0		42.3	10.1		41.6	11.5	

^a The baseline comprised waves 1, 2, and 3 for 7793, 1795, and 283 individuals, respectively.

between low job control and low reward, again suggesting that they overlap with each other.

Table 3 summarizes the results of the regression models that examined the association between physical inactivity and each health-risk factor. It reveals that physical inactivity was associated with most of the health-risk factors. The fixed-effects models (right columns) show that physical inactivity was positively associated with systolic and diastolic blood pressure, triglyceride level, LDL cholesterol level, fasting blood glucose level, HbA1c level, BMI, waist circumference, and K6 scores. On comparison, the pooled cross-sectional models (center columns) failed to show a significant association of physical inactivity with systolic blood pressure, LDL cholesterol level, and HbA1c level at 5% significance, and showed a negative relationship with BMI.

Table 4 shows how the estimated associations between job stress and physical inactivity differed between the pooled cross-sectional and fixed-effects logistic models for JDC (top section) and ERI (bottom section), respectively. The number of respondents in

the fixed-effects logistic models was 3047, which was 30.9% of that in the pooled cross-sectional models, because those who stayed physically active or inactive were removed from the fixed-effects logistic models.

For the JDC model, the pooled cross-sectional models indicated that all job stress types raised the odds of physical inactivity compared to low demands/high control (ie, low strain) jobs. The odds ratio (OR) was highest for high demands/low control (ie, high strain) jobs, followed by low demands/low control (ie, passive) jobs and then high demands/high control (ie, active) jobs. In the fixed-effects logistic models, the association between low demands/low control jobs and physical inactivity became non-significant at 5% significance. Meanwhile, both high demands/low control and high demands/high control jobs remained modestly associated with the increased risk of physical inactivity.

In the ERI model, all types of job stress raised the odds of physical inactivity compared to low effort/high reward jobs in pooled cross-sectional models. However, the OR of low effort/low reward jobs was lower than

Table 2. Key features of job stress: pooled data (N=31 025). [SD=standard deviation.]

	Score			Pairwise correlation ^a with:			Prevalence of physical inactivity (%)		
	Range	Mean	SD	Job demands	Job control	Effort	High (A)	Low (B)	Difference (A-B) ^b
Job demands	12-48	33.0	5.4				61.7	58.9	2.8
Job control	24-96	66.3	9.9	0.210			57.4	63.6	-6.2
Effort	3-12	8.0	1.8	0.617	0.167		61.7	58.3	3.5
Reward	7-28	18.2	3.0	-0.106	0.345	-0.114	57.9	63	-5.1

^a All significantly (P<0.001) different from 0.

^b All significantly (P<0.001) different from 0.

Table 3. The estimated association between physical inactivity and health-risk factors: pooled cross-sectional and fixed-effects models. Controlled for gender, educational attainment, hours worked, income, job types, age, and waves. [SE=standard error; HDL=high density lipoprotein; LDL=low density lipoprotein; HbA1c=hemoglobin A1c; BMI=body mass index]

	Pooled cross-sectional			Fixed-effects		
	Difference ^a /mean %	SE/mean %	Observations N	Difference ^a /mean %	SE/mean %	Observations N
Systolic blood pressure (mmHg)	0.25 ^b	0.14	29 140	0.38 ^c	0.17	9708
Diastolic blood pressure (mmHg)	0.48 ^d	0.17	29 141	0.55 ^d	0.21	9708
Triglyceride level (mg/dL)	3.74 ^e	1.06	26 584	2.98 ^c	1.36	9192
Total cholesterol level (mg/dL)	0.24	0.33	10 105	0.45	0.36	4261
HDL cholesterol level (mg/dL)	-2.46 ^e	0.30	27 175	-0.36	0.24	9430
LDL cholesterol level (mg/dL)	0.56 ^b	0.31	27 062	0.86 ^d	0.30	9337
Fasting blood glucose level (mg/dL)	1.27 ^e	0.28	15 590	0.78 ^c	0.32	5744
HbA1c level (%)	0.25	0.16	23 759	0.57 ^e	0.17	8404
BMI (kg/m ²)	-0.81 ^d	0.27	12 373	0.37 ^d	0.11	6103
Waist circumference (cm)	0.80 ^e	0.15	22 333	0.42 ^e	0.08	8507
K6 scores	11.19 ^e	1.04	30 974	8.87 ^e	1.22	9744

^a Difference in the value of each health-risk factor for those with physical inactivity compared to others, in terms of the ratio to the sample mean.

^b Marginally significantly (P<0.1) different from 0.

^c Significantly (P<0.05) different from 0.

^d Significantly (P<0.01) different from 0.

^e Significantly (P<0.001) different from 0.

Table 4. Estimated association between job stress and physical inactivity: pooled logistic and fixed-effects logistic models. ^a [OR=odds ratio, 95% CI=95% confidence interval]

	Pooled cross-sectional (N=31 025 / 9871)		Fixed-effects (N=10 135 / 3047)	
	OR	95% CI	OR	95% CI
Job demands–control models				
Low demands/high control (low strain)	1.00		1.00	
High demands/high control (active)	1.12 ^c	1.05–1.20	1.17 ^d	1.02–1.34
Low demands/low control (passive)	1.20 ^c	1.11–1.28	1.09	0.93–1.28
High demands/low control (high strain)	1.33 ^c	1.24–1.43	1.22 ^d	1.03–1.43
ERI models				
Low effort/high reward	1.00		1.00	
High effort/high reward	1.19 ^c	1.12–1.27	1.24 ^e	1.07–1.43
Low effort/low reward	1.16 ^c	1.07–1.25	0.99	0.84–1.17
High effort/low reward	1.29 ^c	1.21–1.38	1.28 ^e	1.10–1.50

^a Controlled for gender, educational attainment, hours worked, income, job types, age, and waves.

^b Number observations / number individuals.

^c Significantly (P<0.001) different from 1.00.

^d Significantly (P<0.05) different from 1.00.

^e Significantly (P<0.01) different from 1.00.

those of the other two types of job categories. Meanwhile, fixed-effects logistic models show that low effort/low reward jobs had no significant association with physical inactivity, while both high effort/low reward and high effort/high reward jobs remained modestly associated with the increased risk of physical inactivity. This pattern of results was similar to that for the JDC model, with job demands and control replaced by effort and reward, respectively. This probably reflects overlapping aspects between job demands and effort, between job control and reward, and correspondingly between the JDC and ERI models, as already suggested by table 2.

Finally, table 5 provides the results of the employed ordered logistic models to predict a 4-point categorical variable of physical inactivity. The number of respondents was larger than that in the logistic model presented in table 4 because fewer respondents stayed in one of four physical activity statuses as compared to those who exhibited one of two physical activity levels (active or inactive). The OR in this table indicates how each job stress raised the odds of a reduction in the physical activity level, assuming that the OR was the same across the levels of physical inactivity. Evidently, these results are similar to those presented in table 4, in terms of OR magnitude for each type of job stress, as well as regarding its order and statistical significance, indicating the robustness of the results obtained from the fixed-effects logistic models.

Table 5. Estimated association between job stress and physical inactivity (measured on a four-point scale): pooled ordered logistic and fixed-effects ordered logistic models. ^a [OR=odds ratio, 95% CI=95% confidence interval]

	Pooled cross-sectional (N=31 025 / 9871)		Fixed-effects (N=18 459 / 4178)	
	OR	95% CI	OR	95% CI
Job demands–control models				
Low demands/high control (low strain)	1.00		1.00	
High demands/high control (active)	1.13 ^c	1.06–1.21	1.19 ^d	1.06–1.35
Low demands/low control (passive)	1.20 ^c	1.12–1.28	1.13	0.98–1.30
High demands/low control (high strain)	1.35 ^c	1.26–1.45	1.28 ^c	1.11–1.49
ERI models				
Low effort/high reward	1.00		1.00	
High effort/high reward	1.20 ^c	1.13–1.28	1.27 ^c	1.11–1.44
Low effort/low reward	1.15 ^c	1.07–1.23	1.01	0.87–1.17
High effort/low reward	1.29 ^c	1.21–1.38	1.34 ^c	1.17–1.55

^a Controlled for gender, educational attainment, hours worked, income, job types, age, and waves.

^b Number observations / number individuals.

^c Significantly (P<0.001) different from 1.00.

^d Significantly (P<0.01) different from 1.00.

Discussion

In the current study, we investigated the association between job stress and physical inactivity based on the pooled cross-sectional and fixed-effects models. To the best of our knowledge, the current study is one of the first attempts to address this issue in a non-Western country, ie, Japan, in which the prevalence of physical inactivity was higher than that in many other countries (40).

The results obtained from our pooled cross-sectional analysis showed that job stress—especially, high job strain and ERI—was modestly associated with an increased risk of physical inactivity in leisure time. This adds to evidence supportive of a positive association between job stress and physical inactivity, which has often been observed in previous studies (5–12). Unlike the present study, Tsutsumi et al's study (19) found less significant and inconsistent associations between job characteristics and physical activity outside work, using Japanese data. As pointed out by Tsutsumi et al, it was probably because their index of physical activity outside work did not explicitly measure the intensity of leisure-time physical activity.

A more noticeable finding from the current study was that job stress was associated with increased risk of physical inactivity, even after controlling for individual time-invariant attributes in the fixed-effects logistic models. This result was also supported by the ordered fixed-effects logistic models, which confirmed that

job stress was associated with a lower level of physical activity. These results indicate that the association between job stress and physical inactivity, albeit modest, was real, even if it tended to be overestimated in the cross-sectional models. Our findings were in line with those found in Kouvonon et al's work (12), which applied fixed-effects analysis to Finnish data.

Different results between the cross-sectional and fixed-effects models underscore the importance of controlling for individual time-invariant attributes when assessing the association between job stress and physical inactivity. Specifically, we found that the association between low demands/low control (passive) jobs and physical inactivity became non-significant in the fixed-effects model. It might be possible that individuals who preferred physical inactivity may be inactive by nature, and hence tended to choose low demands/low control jobs in the workplace. Results suggest that without controlling for such an individual attribute, we may overestimate the association between low demands/low control jobs and physical inactivity. This also seems to be true of low effort/low reward jobs, whose association with physical inactivity also became non-significant in the fixed-effects models. In addition to these key findings, we found that physical inactivity was generally associated with increased health risks, which were measured by blood pressure, cholesterol levels, BMI, and other health-risk indicators (see table 1). With these observations combined, we can tentatively argue that physical inactivity mediates the impact of job stress on workers' health. It should be noted, however, that reverse causality cannot be ruled out based on the performed analyses, and the mediation effect was not explicitly tested in the present study. These issues remain to be addressed in future research.

Another noticeable finding was that high job demands (in the JDC model) and high effort (in the ERI model) in terms of the association with physical inactivity, in line with previous studies (10, 11). As seen in tables 2 and 3, high job demands raised the odds of physical inactivity, whether combined with high or low control. Similarly, high effort had a positive association with physical inactivity, whether combined with high or low reward. Meanwhile, low control and low reward are shown to have somewhat amplified the association of physical inactivity with high demands and high effort, respectively.

We acknowledge that the current study has several limitations. First, it should be noted that the fixed-effects models concentrated exclusively on within-individual variations over time. Especially in the case of the fixed-effects logit models, more than a half of the respondents who stayed active or inactive throughout waves were dropped from the analysis. This implies that the observed association between job stress and physical inactivity in the fixed-effects models – which was found

to be attenuated from the pooled cross-sectional models – may be still overstated.

Second, we focused on leisure-time physical inactivity, thereby ignoring other domains of physical activity such as worktime and commuting ones. Previous studies have found that different domains of physical activity as well as their combination were differently associated with workers' health (41-43), suggesting that we have to expand the fixed-effects analysis into the examination of the association of job stress with various domains of physical activity or their combination.

Third, the current study did not identify the one-way causality from job stress to physical inactivity, even though it controlled for individual time-invariant attributes, which likely confounded their association. In particular, it is possible that the occurrence of a health problem makes employees feel that their job is more stressful and, at the same time, spend less time doing physical activity, making it difficult to determine whether the cause of physical inactivity is the health problem or the stress of work. In addition, a reverse causation from physical inactivity to job stress cannot be ruled out. A further study is required to identify causation.

In addition to these key limitations, the following points should be noted in interpreting the estimation results in the present study. First, the models did not control for unobserved variables that were time-variant even though they did control for time-invariant variables, and they did not examine how job stress duration or repeated exposure was associated with physical inactivity. Second, our study sample was not randomly selected and was heavily biased towards male workers. Hence, the results may not apply to the general Japanese working population. Finally, results were not free from biases owing to the exclusion of respondents with missing data and attrition.

Despite these limitations, we can conclude that job stress – especially in the forms of high job strain and ERI – is modestly associated with physical inactivity in leisure time. Given that physical inactivity is closely related to many health-risk factors, as confirmed in our dataset, the results of the current study underscore the need for policy efforts to reduce workplace stress exposure and support physical activity among workers experiencing job stress.

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