# **RESEARCH ARTICLE**

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# Association between insulin resistance and lung function trajectory over 4 years in South Korea: community-based prospective cohort

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# Abstract

**Background:** Hyperglycemic conditions are associated with respiratory dysfunction. Although several studies have reported that insulin resistance (IR) is related to decreased lung function, the association between IR and change in lung function has been rarely studied. This study aimed to investigate the potential association of IR on annual change in lung function using a community-based prospective cohort in Korea.

**Methods:** We selected 4827 Korean participants whose serial lung functions were assessed over 4 years using 1:3 propensity score matching. Exposure was baseline IR estimated with homeostatic model assessment (HOMA-IR), and outcomes were annual changes in lung function determined by calculating the regression coefficient using least-square linear regression analysis.

**Results:** In the multivariate linear regression, per one unit increased log transformed HOMA-IR was associated with decline in FEV<sub>1</sub>%-predicted ( $\beta$ : – 0.23, 95% CI: – 0.36 to – 0.11) and FVC %-predicted ( $\beta$ : – 0.20, 95% CI: – 0.33 to – 0.08), respectively. In the generalized additive model plot, HOMA-IR showed a negative linear association with annual changes in FEV<sub>1</sub>%-predicted and FVC %-predicted. The suggested threshold of HOMA-IR for decline in lung function was 1.0 unit for annual change in FEV<sub>1</sub>%-predicted and 2.2 unit for annual change in FVC %-predicted. Age showed statistically significant effect modification on the relationship between HOMA-IR and annual change in FEV<sub>1</sub>%-predicted. Increased HOMA-IR was associated with the decreased annual change in FEV<sub>1</sub>%-predicted, particularly in older people.

**Conclusions:** In South Korea, increased HOMA-IR was associated with decline in lung function. Since IR was related to decline in FEV<sub>1</sub>%-predicted, particularly in older people, tailored approaches are needed in these populations. The potential pulmonary hazard of IR needs to be confirmed in future studies.

Keywords: Diabetes mellitus, Lung, Insulin resistance, Spirometry

# Background

Diabetes mellitus is a prevalent chronic disease in modern society, with many experiencing diabetes and its complications [1, 2]. Lung function decline has been considered a major complication of diabetes [3–5]. Several studies have found that diabetes and hyperglycemia are associated with the development of various pulmonary

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diseases, including chronic obstructive pulmonary disease, asthma, and interstitial lung disease [6–8]. Therefore, good control of diabetes may prevent future lung diseases.

Increased insulin resistance (IR) has been associated with the future development of obesity, diabetes [9], cardiovascular diseases [10], neurologic impairment [11], and kidney dysfunction [12]. Although several studies have reported the potential association between IR and decreased lung function, none of these studies have evaluated the longitudinal effect of IR on future lung function changes. The aging process can cause diverse physiological, immunological and structural changes in the respiratory system [13, 14]. Thus, analyzing consecutive measurements of lung function may help to interpret the effects of aging on IR and lung function changes. In addition, it can clarify a causal association between IR and lung function.

Since diabetes is associated with future decline in lung function and IR is a fundamental pre-condition of diabetes, we hypothesized that increased IR may be associated with future decline in lung function. Therefore, this study aimed to investigate the association between IR and lung function change using data from a community-based prospective Ansan-Ansung cohort in Korea.

# Methods

## Participants

The Ansan-Ansung cohort comprise participants aged 40-69 years who lived in Ansan (urban) and Ansung (rural) areas, and this cohort was analyzed to determine the factors affecting the incidence of chronic diseases. Baseline measurements were performed between May 2001 and February 2003, and two subsequent lung function tests were performed biennially thereafter. More detailed information about the Ansan-Ansung cohort can be obtained in previous reports [15, 16]. Of the 10,030 participants, we excluded 983 missing baseline measurements. Among the 9047 subjects, 4827 subjects with additional spirometry were selected using 1:3 propensity score matching with 1609 subjects without additional spirometry. The differences in clinical characteristics of participants with or without additional spirometry were described in Additional file 3: Table S1. Therefore, 4827 matched participants were finally included in this analysis (Fig. 1).

# **Exposure: IR**

IR was assessed using the homeostatic model assessment for IR (HOMA-IR). HOMA-IR was calculated using the following equation: fasting insulin ( $\mu$ IU/mL) × fasting glucose (mg/dL)/405 [17].

# **Outcomes: change in lung function**

The main outcomes were annual changes in lung function. Lung function tests were performed using a spirometer (VMAX2130; Sensormedics Corporation, Yorba, CA, USA). Pre-bronchodilator values were measured three times by trained technicians, and the best scores were recorded. Predicted values of functional vital capacity (FVC %-predicted) and functional expiratory volume in 1 s (FEV<sub>1</sub>%-predicted) were calculated using the Korean formula [18]. The FEV<sub>1</sub>/FVC ratio was expressed as the percentage of raw value of FEV<sub>1</sub> divided by the raw value of FVC, according to previous literature [19, 20]. Annual changes in FEV<sub>1</sub>%-predicted, FVC %-predicted, and FEV<sub>1</sub>/FVC ratio was determined by calculating the regression coefficient using least-square linear regression analysis, with FEV<sub>1</sub>%-predicted, FVC %-predicted, and FEV<sub>1</sub>/ FVC ratio as functions of time in years. It was applied to all values of FEV<sub>1</sub>%-predicted, FVC %-predicted, and FEV<sub>1</sub>/FVC ratio that was obtained during the follow-up period (Additional file 2: Figure S1).

# Measurements and other definitions

Trained investigators interviewed participants regarding their socioeconomic status and lifestyle habits using Additional file 1: KoGES Baseline Core Questionnaire. High income was defined as the highest quintile of monthly household income ( $\geq 3$  million won a month). Active physical activity was defined as 60 min/day of moderate-intensity activity or 30 min/day of vigorousintensity activity [21]. Blood pressure (BP) was defined as the average BP on both arms using a standard mercury sphygmomanometer (Baumanometer-Standby; W. A. Baum Co., Inc., Copiague, NY, USA). High BP was defined as systolic BP>140 mmHg or diastolic BP > 90 mmHg. Body mass index (BMI) was defined by dividing the weight by the square of the height  $(kg/m^2)$ . Waist circumference was measured at the narrowest point between the lower rib and the iliac crest (measured to the nearest 0.1 cm). All blood samples were examined after fasting for at least 8 h. Hemoglobin and white blood cell (WBC) count were analyzed using enzymatic methods with ADVIA 120 (Bayer Diagnostics, Tarrytown, NY, USA). Fasting glucose, hemoglobin A1c (HbA1c), triglyceride, high density lipoprotein (HDL) cholesterol, C-reactive protein (CRP), and serum creatinine levels were measured using ADVIA 1650 (Siemens, Tarrytown, NY, USA). Estimated glomerular filtration rate (eGFR) was defined using the Chronic Kidney Disease Epidemiology Collaboration equation [22]. Annual averaged values were obtained for BMI, waist circumference, systolic and diastolic BP, triglyceride, HDL cholesterol, eGFR,



WBC count, fasting glucose, HbA1c, hemoglobin, and CRP level.

# Statistical analyses

We compared individuals without additional spirometry or with missing values to the rest of participants. Variables that showed significant differences were selected for matching. As a result, age, sex, high income, college graduate, smoking status, active physical activity, baseline values of BMI, systolic BP, diastolic BP, WBC, CRP, and FEV<sub>1</sub>%-predicted were used in the propensity score matching. The normality of the distribution of continuous variables was assessed using histogram and Q–Q plots. Mean $\pm$  standard deviation (SD) was used to express normally distributed continuous variables, whereas median with interquartile range (IQR) was used for non-normally distributed continuous variables. The *P*-trend was analyzed using linear regression for the normally distributed continuous variables, the Jonckheere– Terpstra test for the non-normally distributed continuous variables, and the Cochran–Armitage test for categorical variables. We further analyzed the difference between individuals with and without additional spirometry using student's t-test for a normally distributed continuous variable, Wilcoxon rank-sum test for non-normally distributed continuous variable, and Pearson's Chi-squared test for categorical variables. The correlation between age and HOMA-IR was performed using Spearman's rank correlation analysis.

The simple associations between HOMA-IR and spirometric parameters were expressed using the locally weighted scatterplot smoothing (LOWESS) function. We performed the linear regression analysis between log-transformed HOMA-IR and annual changes in  $FEV_1$ %-predicted and FVC %-predicted. Odds ratio (OR) and 95% confidence interval (CI) for the lowest quartile of annual changes in FEV<sub>1</sub>%-predicted and FVC %-predicted were analyzed using logistic regression analysis.

In multivariate analyses, covariates were chosen based on clinical and statistical relevance. The non-linear association between HOMA-IR and annual changes in  $FEV_1$ %-predicted and FVC %-predicted was analyzed using the multivariate generalized additive model (GAM) for Gaussian distributions with the "mgcv" package, and calculation of Akaike information criterion (AIC) was used for the model fitting. In AIC, lower scores within the data set indicate a better model fit [23]. The threshold point of HOMA-IR was chosen based on the best fit determined by AIC among the models.

In the subgroup analysis, all control variables were used, and participants were divided by the median for continuous variable. The modulatory effect was confirmed by adding the interaction term of variables defining the subgroup to the multivariate linear regression analysis. To visualize potential interaction of age on the association between IR and annual change in lung function, the LOWESS regression with interaction was implemented using the "predict3d" package. A *P*-value < 0.05 was considered statistically significant. All statistical analyses were performed using R version 3.6.2 (R Core Team 2019; R Foundation for Statistical Computing, Vienna, Austria).

## Results

The mean age of the 4827 participants was 52.6 years, 45.3% were men, 48.0% had active physical activity, and 41.7% were current or past smokers. The mean  $\pm$  SD of annual averaged BMI, systolic and diastolic BP, waist circumference, HDL cholesterol, eGFR, hemoglobin, and WBC count were  $24.4 \pm 3.0 \text{ kg/m}^2$ ,  $119.6 \pm 15.6 \text{ mmHg}$ ,  $79.0 \pm 9.4$  mmHg,  $83.2 \pm 8.6$  cm,  $44.9 \pm 8.8$  mL/dL,  $82.9 \pm 10.9$  mL/min/1.73m<sup>2</sup>,  $13.6 \pm 1.4$  g/dL, and  $6.5 \pm 1.5 \times 10^3 / \mu$ L, respectively. Moreover, the median (IQR) of annual averaged fasting glucose, HbA1c, triglyceride, and CRP level were 88 (83-94) mg/dL, 5.6 (5.3-5.8) %, 130 (97-178) mg/dL, and 0.12 (0.07-0.22) mg/dL, respectively. At baseline, the mean  $\pm$  SD values of FEV<sub>1</sub>%-predicted, FVC %-predicted, and FEV<sub>1</sub>/ FVC ratio were  $94.9 \pm 14.1\%$  predicted,  $95.4 \pm 13.2\%$ predicted, and  $79.5 \pm 7.9\%$ , respectively. The mean of FEV<sub>1</sub>%-predicted, FVC %-predicted, and FEV<sub>1</sub>/FVC ratio decreased in subsequent lung function measurement (Fig. 2). The mean  $\pm$  SD values of annual changes in FEV<sub>1</sub>%-predicted, FVC %-predicted, and FEV<sub>1</sub>/FVC ratio were  $-0.42 \pm 2.62\%$ -predicted/year,  $-0.49 \pm 2.72\%$  predicted/year, and  $-0.35 \pm 1.56\%$ /year, respectively.



	HOMA-IR groups (n = 4827)						
	1Q:<1.05 (n=1218)	2Q: 1.05–1.45 (n = 1211)	3Q: 1.46-2.03 (n = 1191)	$4Q: \ge 2.04 (n = 1207)$	P-trend		
Age(years)	53.06±9.11	52.53±8.86	52.49±8.84	$52.50 \pm 8.93$	0.131		
Male, n (%)	649 (53.3)	551 (45.5)*	462 (38.8)*†	524 (43.4)*‡	< 0.001		
High income, n (%)	161 (13.2)	170 (14.0)	160 (13.4)	182 (15.1)	0.263		
College graduate, n (%)	119 (9.8)	129 (10.6)	119 (10.0)	127 (10.5)	0.681		
Current or past smoker, n (%)	582 (47.8)	494 (40.8)*	433 (36.4)*†	506 (41.9)*‡	< 0.001		
Active physical activity, n (%)	595 (48.8)	631 (52.1)*	553 (46.4)†	536 (44.4)*†	0.003		
Annual averaged BMI (Kg/m²)	$23.32 \pm 2.95$	$23.83 \pm 2.82^*$	24.64±2.75*†	25.83±3.03*†‡	< 0.001		
Annual averaged systolic BP (mmHg)	$117.72 \pm 15.32$	$117.95 \pm 15.30$	119.49±15.09*	123.44±16.13*†‡	< 0.001		
Annual averaged diastolic BP (mmHg)	$77.65 \pm 9.35$	$78.05 \pm 9.45$	79.07±9.17*†	81.14±9.16*†‡	< 0.001		
Annual averaged waist circumference (cm)	$80.26 \pm 8.22$	81.63±7.97*	83.78±8.25*†	87.18±8.22*†‡	< 0.001		
Annual averaged fasting glucose (mg/dL)	85 (81–90)	86 (82–91)*	88 (83–94)*†	94 (87–109)*†‡	< 0.001		
Annual averaged HbA1c (%)	5.4 (5.3–5.7)	5.5 (5.3–5.8)	5.6 (5.3–5.8)*†	5.8 (5.5–6.4)*†‡	< 0.001		
Annual averaged triglyceride (mg/dL)	118 (90–162)	120 (91–162)	129 (98–175)*†	154 (115–210)*†‡	< 0.001		
Annual averaged HDL cholesterol (mg/dL)	$46.47 \pm 9.25$	$45.65 \pm 8.91$	44.73±8.70*	42.71±8.02*†‡	< 0.001		
Annual averaged eGFR (mL/min/1.73m <sup>2</sup> )	$83.44 \pm 10.98$	$83.40 \pm 10.68$	82.43±10.77*†	$82.15 \pm 11.16* \dagger$	0.004		
Annual averaged hemoglobin (g/dL)	$13.64 \pm 1.41$	$13.51 \pm 1.42$	$13.45 \pm 1.51*$	13.69±1.42†‡	0.010		
Annual averaged WBC count ( $\times 10^{3}/\mu$ L)	$6.40 \pm 1.55$	$6.34 \pm 1.47$	$6.43 \pm 1.60$	6.78±1.52*†‡	< 0.001		
Annual averaged CRP level (mg/dL)	0.10 (0.06–0.20)	0.10 (0.06-0.20)	0.12 (0.07-0.21)*†	0.15(0.08-0.26)*†‡	< 0.001		
Baseline FEV1 (L)	$2.84 \pm 0.68$	$2.81 \pm 0.68$	$2.77 \pm 0.68$	$2.78 \pm 0.68$	0.045		
Baseline FVC (L)	$3.63 \pm 0.86$	$3.57 \pm 0.87$	$3.49 \pm 0.86^{*}$	3.49±0.86*	< 0.001		
Baseline FEV1 (%-predicted)	$94.73 \pm 14.46$	$95.26 \pm 13.87$	$95.60 \pm 14.28$	$94.18 \pm 13.57$	0.071		
Baseline FVC (%-predicted)	$96.26 \pm 13.19$	$96.27 \pm 13.30$	$95.81 \pm 13.61$	93.10±12.33*†‡	< 0.001		
Baseline FEV1/FVC (%)	$78.68 \pm 8.59$	$79.33 \pm 7.90$	79.76±7.82*	$80.06 \pm 7.28^{*}$	< 0.001		

Table 1 Clinical characteristics of the study population according to HOMA-IR quartile

Values are expressed as mean  $\pm$  standard deviation for normally distributed continuous variables, median and interquartile range for non-normally distributed variables and percentage for categorical variables. P-trend was analyzed normally distributed continuous variables by the linear regression, for non-normally distributed distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by the linear regression, for non-normally distributed continuous variables by Lonckheere-Terpstra tests, and for categorical variables by Cochran-Armitage test for trend

\*, †, and ‡ meant P<0.05 when compared to<1.05, 1.06–1.45, 1.45–2.03 groups of HOMA-IR, respectively, using Bonferroni post-hoc analysis of one-way ANOVA for normally distributed continuous variables, Mann–Whitney U tests for non-normally distributed continuous variable, and Chi-square tests for categorical variables

The clinical characteristics of the study population according to HOMA-IR quartile are shown in Table 1. Although age was not associated with HOMA-IR quartile, the increased HOMA-IR quartile was associated with a decreased proportion of men. The correlation analysis between age and HOMA-IR also was not statistically significant (Additional file 2: Fig. S2). Although income and educational status were not associated with HOMA-IR quartile, the increased HOMA-IR quartile was associated with a decreased rate of active physical activity and current or past smoking. With the increase of HOMA-IR quartile, BMI, systolic and diastolic BP, waist circumference, WBC count, fasting glucose, HbA1c, triglyceride, hemoglobin, and CRP level increased, whereas HDL cholesterol decreased. With the increase of HOMA-IR quartile, baseline FVC %-predicted decreased, but FEV<sub>1</sub>/FVC increased. Baseline FEV<sub>1</sub>%-predicted was not associated with HOMA-IR.

We explored the association between HOMA-IR and annual changes in lung function (Fig. 3). As HOMA-IR increased, annual changes in FEV<sub>1</sub>%-predicted and FVC %-predicted change decreased, whereas annual change in FEV<sub>1</sub>/FVC ratio change increased. In multivariate linear regression analysis, per one unit increased log transformed HOMA-IR was associated with decline in FEV<sub>1</sub>%-predicted ( $\beta$ : - 0.23, 95% CI: - 0.36 to - 0.11, p < 0.001) and FVC %-predicted ( $\beta$ : – 0.20, 95% CI: - 0.33 to - 0.08, p = 0.001) (Table 2). In the categorical analysis, the odds for the lowest quartile of annual changes in FEV<sub>1</sub>%-predicted (Q1: < -1.73%-predicted/ year) and FVC %-predicted (Q1: - 1.90%-predicted/year) showed increasing trend (P-trend < 0.001) as quartiles of HOMA-IR were increased. The OR of HOMA-IR for lowest quartile of annual changes in FEV1%-predicted and FVC %-predicted were significantly higher in the upper quartiles compared to the lower quartile. In the



multivariate GAM plot analysis, we identified that the association between HOMA-IR and annual changes in FEV<sub>1</sub>%-predicted and FVC %-predicted were negative linear, and the lowest AICs were found in HOMA-IR 1.0 unit for annual change in FEV<sub>1</sub>%-predicted and 2.2 unit for annual change in FVC %-predicted (Fig. 4).

There were no subgroups showing the effect modification between HOMA-IR and annual changes in FVC % predicted (Additional file 3: Table S3). However, participant age significantly modified the effect of HOMA-IR on annual change in FEV<sub>1</sub>%-predicted (*p* for interaction = 0.003). Although HOMA-IR was not associated with annual change in FEV<sub>1</sub>%-predicted in people aged < 50 years, increased HOMA-IR was significantly associated with the decreased annual change in FEV<sub>1</sub>%-predicted in FEV<sub>1</sub>%-predicted in individuals aged  $\geq$  50 years (Table 3). We also identified similar effect modification of age on the association between HOMA-IR and annual change in FEV1%-predicted in the LOWESS regression plot with interaction (Fig. 5).

# Discussion

Our study revealed that increased IR associated with decline in lung function using community-based prospective cohort in Korea. Recently, decreased lung function has been suggested as a complication of IR. In their large-scale cross-sectional study, Lawlor et al. found that HOMA-IR decreased by 5% and 8% for a 1 SD increase in log  $FEV_1$  and FVC, respectively [24]. Sagun et al. also showed that decreased FEV<sub>1</sub>%-predicted was correlated with increased insulin resistance (HOMA-IR  $\geq$  2.5) in an age-adjusted analysis (OR 0.97, 95% CI 0.94-1.00, P=0.028) [25]. However, most previous studies were cross-sectional, and a causal association between insulin resistance and decreased lung function needed to be evaluated in a prospective cohort. Therefore, we conducted the current study using a prospective communitybased cohort and found that increased insulin resistance was associated with a decline in lung function, particularly in the elderly population.

Table 2	The effects	of HOMA-IR o	n the lung	function	change

Exposure	Outcom	Outcome: Annual lung function change					
	Δ FEV1 (9	Δ FEV1 (%-predicted/year) Beta (95% Cl, P)		∆ FVC (%-predicted/year) Beta (95% Cl, P))			
Log transformed HOMA-IR	Beta (959						
Per one unit increase (n = 4827)	- 0.23 (-	- 0.23 (- 0.36 to - 0.11, < 0.001)		- 0.20 (- 0.33 to - 0.08, 0.001)			
HOMA-IR quartile	Lowest quartile (Q1) of ∆ FEV 1.73%-predicted/year	1:<-	P-trend	Lowest quartile (Q1) of $\Delta$ FVC: < – 1.90%-predicted/year	P-trend		
	Adjusted OR (95% CI, P)			Adjusted OR (95% Cl, P)			
1Q:<1.05 (n=1218)	1 (reference)		< 0.001	1 (reference)	< 0.001		
2Q: 1.05–1.44 (n = 1211)	1.14 (0.94–1.39, 0.191)			1.15 (0.94–1.41, 0.162)			
3Q: 1.45-2.02 (n = 1191)	1.22 (1.00–1.48, 0.055)			1.30 (1.06–1.59, 0.012)			
4Q:≥2.03 (n=1207)	1.34 (1.09–1.66, 0.006)			1.38 (1.11–1.71, 0.004)			

The rapid decline in lung function was defined in two aspects, which were the first quartile of FEV<sub>1</sub>%-predicted and FVC %-predicted, respectively. ORs were adjusted for age, sex, college graduate, high income, smoking status, active physical activity, and annual averaged values of BMI, systolic and diastolic BP, waist circumference, fasting glucose, HbA1c, triglyceride, HDL cholesterol, eGFR, WBC count, hemoglobin, and CRP level. *P*-trend for HOMA-IR quartile was analyzed by Cochran-Armitage test for trend

Exposure: per one unit increase	Adjusted beta (CI, P-value)	e) Outcome: annual change in lung function				
log-transformed HOIMA-IR		Δ FEV1 (%-predicted/year)		∆ FVC (%-predicted/year)		
Subgroup	No. of people	P for interaction	Adjusted beta (Cl, <i>P</i> -value)	P for interaction		
Age (years)	< 50 (n = 2177)	0.04 (- 0.13 to 0.22, 0.613)	0.003	0.03 (- 0.15 to 0.20, 0.775)	0.153	
	$\geq$ 50 (n = 2650)	- 0.41 (- 0.58 to - 0.24, < 0.001)		- 0.33 (- 0.51 to - 0.16, < 0.001)		
Sex	Male (n = 2186)	- 0.24 (- 0.41 to - 0.07, 0.006)	0.798	- 0.15 (- 0.33 to 0.03, 0.103)	0.554	
	Female (n = 2641)	- 0.22 (- 0.40 to - 0.05, 0.012)		- 0.24 (- 0.41 to - 0.06, 0.008)		
Current or past smoker	No (n $=$ 2812)	- 0.18 (- 0.34 to - 0.02, 0.027)	0.447	- 0.18 (- 0.34 to - 0.02, 0.012)	0.941	
	Yes (n = 2015)	- 0.30 (- 0.48 to - 0.11, 0.002)		- 0.18 (- 0.37 to 0.01, 0.060)		
Active physical activity	No (n $= 2512$ )	- 0.17 (- 0.32 to - 0.02, 0.028)	0.496	- 0.12 (- 0.28 to 0.04, 0.002)	0.423	
	Yes (n = 2315)	- 0.31 (- 0.50 to - 0.11, 0.002)		- 0.29 (- 0.49 to - 0.10, 0.004)		
High BP	No (n = 4043)	- 0.21 (- 0.35 to - 0.08, 0.002)	0.579	- 0.20 (- 0.33 to - 0.06, 0.004)	0.919	
	Yes (n = 784)	- 0.35 (- 0.66 to - 0.03, 0.030)		- 0.27 (- 0.60 to 0.06, 0.104)		
Annual averaged BMI (kg/m²)	< 24 (n = 2223)	- 0.30 (- 0.50 to - 0.11, 0.002)	0.345	- 0.22 (- 0.42 to - 0.03, 0.023)	0.998	
	$\geq$ 24 (n = 2604)	- 0.17 (- 0.33 to - 0.02, 0.031)		- 0.19 (- 0.35 to - 0.02, 0.026)		
Annual averaged eGFR (mL/ min/1.73m <sup>2</sup> )	< 84 (n = 2496)	- 0.22 (- 0.39 to - 0.05, 0.010)	0.387	- 0.10 (- 0.27 to 0.07, 0.256)	0.751	
	$\geq$ 84 (n = 2331)	- 0.24 (- 0.42 to - 0.06, 0.010)		- 0.31 (- 0.49 to - 0.12, 0.001)		
Annual averaged WBC count (× 10³/µL)	< 6.3 (n = 2396)	- 0.26 (- 0.44 to - 0.08, 0.006)	0.643	- 0.15 (- 0.34 to 0.04, 0.121)	0.735	
	$\geq$ 6.3 (n = 2431)	- 0.21 (- 0.37 to - 0.05, 0.012)		- 0.22 (- 0.40 to - 0.05, 0.013)		
Annual averaged Hemoglobin (mg/dL)	< 13.5 (n = 2462)	- 0.17 (- 0.35 to 0.02, 0.080)	0.713	- 0.12 (- 0.31 to 0.07, 0.217)	0.425	
	$\geq$ 13.5 (n = 2365)	- 0.29 (- 0.45 to - 0.14, < 0.001)		- 0.24 (- 0.40 to - 0.08, 0.004)		

 Table 3
 Subgroup analysis for the effects of HOMA-IR on annual change in lung function

Adjusted beta and 95% CIs were analyzed using the multivariate linear regression. Age, sex, college graduate, high income, smoking status, active physical activity, and annual averaged values of BMI, systolic and diastolic BP, waist circumference, fasting glucose, HbA1c, triglyceride, HDL cholesterol, eGFR, WBC count, hemoglobin, and CRP level were included for adjustment. The variable used to divide subgroup was excluded from this analysis

In the current study, increased HOMA-IR was independently associated with decline in FEV<sub>1</sub>%-predicted and FVC %-predicted. Cigarette smoking is a well-established risk factor for lung function deterioration. In the large scale randomized clinical trial, annual change in  $FEV_1$ %-predicted was approximately - 1.4%-predicted/ year in the continuing smoker group [26]. Compared to this result, IR can be considered a risk factor for rapid decline in lung function. In LOWESS regression analysis, increased HOMA-IR presented negative association with annual changes in FEV<sub>1</sub>%-predicted and FVC %-predicted, but not with FEV<sub>1</sub>/FVC ratio. In the natural history of chronic obstructive pulmonary disease, the rate of  $FEV_1$  decline is slow at high  $FEV_1$  and faster as  $FEV_1$  decreases [27]. Due to the nature of the community-based cohort, the majority of participants had a high baseline FEV<sub>1</sub>. These factors could explain why HOMA-IR was not associated with annual change in FEV<sub>1</sub>/FVC ratio.

We also found negative linear associations between HOMA-IR and annual changes in  $FEV_1$ %-predicted

and FVC %-predicted using the multivariate GAM plot, and the suggested threshold of HOMA-IR for decline in lung function was 1.0 unit for annual change in FEV<sub>1</sub>%-predicted and 2.2 unit for annual change in FVC %-predicted. As a surrogate marker of IR, several studies have shown that the cutoff value of HOMA-IR for the risk of developing type 2 diabetes mellitus and metabolic syndrome in the Korean population is approximately 2.5 [28–30]. Our findings showed lower threshold than those of previous studies, indicating that decline in lung function may be more sensitive to the systemic effects of IR. Moreover, these results suggest that early treatment of IR can be an effective intervention to prevent decline in lung function.

There were two possible explanations for the potential pulmonary hazard of IR. First, chronic low-grade inflammation in adipose tissue contributes to the development of IR [31, 32]. Levels of cytokines, such as tumor necrosis factor-alpha (TNF- $\alpha$ ) and interleukin-6, are increased with IR and cause lung fibrosis and inflammation [33, 34]. Second, respiratory muscle weakness is related to IR.



In high IR status, TNF- $\alpha$  and interferon-gamma cause muscle wasting through the degeneration of myotubes and modulation of myogenesis [35, 36].

In subgroup analysis, age showed a statistically significant effect modification on the association between HOMA-IR and annual change in  $FEV_1$ %-predicted. Although HOMA-IR was not associated with annual change in  $FEV_1$ %-predicted in the younger group, increased HOMA-IR was significantly related to decline in  $FEV_1$ %-predicted in the older group.

IR is associated with chronic low-grade inflammation [31]. The risk of chronic low-grade inflammation increased with the aging process [37, 38]. Lung function can decline with increased inflammation [39, 40]. Therefore, we assumed that increased chronic low-grade inflammation during the aging process may potentiate





the metabolic hazard of IR in lung function deterioration. In this regard, early intervention to reduce IR in the elderly population may have a beneficial effect on future lung function preservation.

Several limitations should be considered when interpreting our findings. First, two or fewer additional lung function measurements were performed in this study. Therefore, our study period was not sufficiently long. Further studies performing more spirometric tests will be needed to clarify the long-term effect of IR on decline in lung function. Second, our study was conducted in single ethnicity and selected areas. Therefore, the results should be elucidated cautiously in accordance with racial and regional differences.

Despite these limitations, our study has several strengths. First, to the best of our knowledge, this is the

first study to investigate the association between IR and lung function using a large-scale prospective cohort based on the general population. Second, we calculated the threshold of HOMA-IR for decline in lung function using a nonlinear analytic method. Finally, the interaction between age and IR to annual change in FEV<sub>1</sub>%-predicted was investigated and presented using visualization.

# Conclusion

In South Korea, increased IR was independently associated with decline in FEV<sub>1</sub>%-predicted and FVC %-predicted. We also found that age modified the relationship between IR and FEV<sub>1</sub>%-predicted, particularly valid in older people. The potential pulmonary hazard of IR needs to be confirmed in future interventional studies.

## Abbreviations

HOMA-IR: Homeostatic model assessment-insulin resistance; FEV1: Forced expiratory volume in 1 s; FVC: Functional vital capacity; OR: Odds ratio; CI: Confidence interval; BP: Blood pressure; BMI: Body mass index; WBC: White blood cell; HbA1c: Hemoglobin A<sub>1</sub>c; HDL: High density lipoprotein; CRP: C-reactive protein; eGFR: Estimated glomerular filtration rate; SD: Standard deviation; IQR: Interquartile range; LOWESS: Locally weighted scatter-plot smoothing; GAM: Generalized additive model; AIC: Akaike's information criterion; TNF- $\alpha$ : Tumor necrosis factor-alpha.

# Supplementary Information

The online version contains supplementary material available at https://doi.org/10.1186/s12890-021-01478-7.

Additional file 1. KoGES baseline Core Questionnaire.

Additional file 2: Figure S1. Definitions of outcome and exposure variables. Abbreviations: HOMA-IR, homeostatic model assessment-insulin resistance. Figure S2. Correlation between age and HOMA-IR. Abbreviations: HOMA-IR, homeostatic model assessment-insulin resistance.

Additional file 3: Tables S1. Comparison of participants' baseline characteristics with or without additional spirometry and Table S2. Additional subgroup analysis for the effects of HOMA-IR on annual change in lung function.

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#### Authors' contributions

S.W.L. conceptualization. S.H.K. data curation/formal analysis. H.S.K. investigation/ methodology. S.H.K. writing – original draft. H.K.M. and H.S.K. writing – review and editing. S.W.L. supervision/validation. All authors read and approved the final manuscript.

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#### Availability of data and materials

The data of our study is fully available when manuscript is accepted for publication.

# Declarations

#### Ethics approval and consent to participate

Using the Ansan-Ansung cohort, this study was conducted according to the Declaration of Helsinki. The institutional review boards of the Nowon Eulji Medical Center, Eulji University approved the protocol and ethics of the study (Approval Number: 2019-06-014). Participants provided written informed consent, and all data were completely anonymized prior to access.

#### **Consent for publication**

Not applicable.

#### Competing interests

All authors have no conflict of interest to declare.

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