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# Association of fasting plasma glucose variability with gestational diabetes mellitus: a nationwide populationbased cohort study

Jung A Kim,<sup>1</sup> Jinsil Kim,<sup>2</sup> Eun Roh,<sup>1</sup> So-hyeon Hong,<sup>1</sup> You-Bin Lee,<sup>1</sup> Sei Hyun Baik,<sup>1</sup> Kyung Mook Choi <sup>(i)</sup>, <sup>1</sup> Eunjin Noh,<sup>2</sup> Soon Young Hwang,<sup>2</sup> Geum Joon Cho,<sup>3</sup> Hye Jin Yoo <sup>(i)</sup>

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<sup>1</sup>Division of Endocrinology and Metabolism, Department of Internal Medicine, College of Medicine, Korea University, Seoul, Republic of Korea <sup>2</sup>Smart Healthcare Cancer, Korea University Guro Hospital, Seoul, Republic of Korea <sup>3</sup>Department of Obstetrics and Gynecology, Korea University Guro Hospital, Korea University College of Medicine, Seoul, Republic of Korea

#### **Correspondence to**

Dr Hye Jin Yoo; deisy21@naver.com and Dr Geum Joon Cho; md cho@hanmail.net

#### ABSTRACT

**Objective** Long-term glycemic variability has recently been recognized as another risk factor for future adverse health outcomes. We aimed to evaluate the risk of gestational diabetes mellitus (GDM) according to the prepregnancy long-term fasting plasma glucose (FPG) variability.

**Research design and methods** A total of 164053 women who delivered their first baby between January 1, 2012 and December 31, 2015, were selected from the Korean National Health Insurance data. All women underwent at least three national health screening examinations, and the last examination should be conducted within 2 years before their first delivery. GDM was defined as the presence of more than four times of claim of GDM (International Classification of Disease, 10th Revision (ICD-10) 024.4 and 024.9) or prescription of insulin under the ICD-code of GDM. FPG variability was assessed by variability independent of the mean (FPG-VIM), coefficient of variation, SD, and average successive variability.

**Results** Among the 164053 women, GDM developed in 6627 (4.04%). Those in the higher quartiles of FPG-VIM showed a stepwise increased risk of GDM. In fully adjusted model, the ORs for GDM was 1.22 (95% Cl 1.14 to 1.31) in women with the highest FPG-VIM quartile compared with those in the lowest quartile. The risk for GDM requiring insulin therapy was 48% increase in women in the highest quartile of FPG-VIM compared with those in the lowest quartile, while that for GDM not requiring insulin therapy was 19% increase. The association between high FPG variability and the risk of GDM was intensified in the obese and aged more than 35 years women.

**Conclusions** Increased FPG variability in the prepregnancy state is associated with the risk of GDM independent of confounding factors. Therefore, prepregnancy FPG variability might be a surrogate marker of the risk of GDM.

#### INTRODUCTION

Parallel with the increasing prevalence of diabetes, obesity, and the late age of pregnancy, gestational diabetes mellitus (GDM) has become one of the most prevalent

#### Significance of this study

#### What is already known about this subject?

 Long-term glycemic variability has recently been recognized as a risk factor for future adverse health outcomes.

Original research

Although gestational diabetes mellitus (GDM) affects the mother and their offspring during the perinatal period and the life-long period, it is hard to find out GDM in early stage in primipara.

#### What are the new findings?

- Among the 164 053 women, GDM developed in 6627 (4.04%) and women in the higher quartiles of fasting plasma glucose (FPG)-variability independent of the mean (VIM) showed a stepwise increased risk of GDM.
- The risk for GDM was 22% in the highest quartile of FPG-VIM compared with those in the lowest quartile after adjusting risk factors.
- The association between high FPG variability and the risk of GDM was intensified in the obese and aged more than 35 years.

## How might these results change the focus of research or clinical practice?

The early risk stratification of GDM using prepregnancy FPG variability might be clinically valuable, especially for the high-risk women such as obese, aged women.

pregnancy-related complications.<sup>1</sup> GDM is associated with pre-eclampsia, preterm delivery, fetal overgrowth, neonatal hypoglycemia, and increased mortality in the perinatal period.<sup>2</sup> Furthermore, women with previous GDM and offspring exposed to maternal hyperglycemia have an increased risk of obesity, type 2 diabetes (T2DM), and cardiometabolic diseases.<sup>3–5</sup> As a result, GDM causes a continuous intergenerational transfer of metabolic disturbances; thus, early identification of women at risk for GDM and intensive strategies for reducing risk factors is critical to break this vicious cycle. However, there has been considerable controversy regarding the early screening method for GDM.

Recent growing evidences have emphasized the role of variability of certain metabolic parameters in predicting the risk for various adverse health outcomes.<sup>6–8</sup> Representatively, Kim *et al*<sup> $\overline{l}$ </sup> found a graded association between the number of high-variability parameters including fasting plasma glucose (FPG) and total cholesterol (TC) concentrations, systolic blood pressure (BP), and body mass index (BMI) and primary outcomes such as mortality and cardiovascular events. Long-term glycemic variability, usually demonstrated as a visit-to-visit FPG or HbA1C variability, indicates glucose fluctuation over months to vears.<sup>9</sup> The excess glycemic excursion associated with increased oxidative stress, endothelial dysfunction, and subclinical inflammation promotes organ damage.<sup>10</sup> In human umbilical vein endothelial cells, glucose oscillation induces more oxidative stress than consistent hyperglycemia, resulting in the acceleration of apoptosis and inflammatory reactions.<sup>11</sup> Although previous studies have extensively investigated the intimate relationship between glycemic variability and micro/macrovascular complications and overall mortality in patients with diabetes,<sup>12 13</sup> only a few have examined the prognostic value of long-term glycemic variability in individuals without diabetes. Recently, Wang et al<sup>14</sup> showed that FPG variability over 4 years was an independent predictor of cardiovascular diseases (CVD) and all-cause mortality in the general population. Bancks *et al*<sup>15</sup> also reported that greater FPG variability during young adulthood was associated with higher risk for incident diabetes, CVD, and all-cause mortality during the middle age years. In a nationwide population study including subjects aged  $\geq 40$ years without diabetes, a 1 SD increase in the FPG variability was associated with 24% increased risk of T2DM within the 8.3-year follow-up.<sup>16</sup> However, no previous study has explored association of prepregnancy longterm FPG variability on the risk of GDM, especially to the primipara with baseline normal fasting glucose level, as GDM is difficult to detect during the early stage.

Therefore, to search for a novel risk factor for GDM, we aimed to investigate the association between prepregnancy FPG variability and the risk of GDM in the primipara using the large-scale Korean National Health Insurance Service (NHIS) database.

#### MATERIALS AND METHODS Data source

The data were obtained from the NHIS database from 2002 to 2015. The NHIS is operated by the Korean government, and all enrolled participants are required to undergo annual or biannual health examinations. The NHIS database is composed of anonymous personal information, health examination database, demographics, primary and secondary diagnoses, hospitalization, and

medical treatment. In this study, 1625913 women who delivered their first baby between January 1, 2012 and December 31, 2015, who underwent health examinations within 2 years before delivery (index year), and who underwent additional health examinations ( $\geq 2$ ) within 5 years before the last health check-up (index year) were included (n=180094). We excluded women who had multiple births (n=27266), multiparity(n=789033), FPG≥100 mg/dL at index year (n=15582), a history of diabetes before index year (n=19), and had at least one missing data in BMI, hemoglobin, weight, BP, smoking, drinking, and exercise (n=440) (figure 1). In total, 164053 women were eligible for the analysis. Women with GDM were identified using the International Classification of Disease, 10th Revision (ICD-10) (O24.4 and O24.9). Categorically, 157 426, 5808, and 819 women were subgrouped as follows: normal pregnancy, GDM without insulin therapy, and GDM with insulin therapy. Our study protocol was approved by the Korea University institutional review board (IRB) and was conducted in accordance with the Declaration of Helsinki of the World Medical Association (2019GR0106). An informed consent exemption was granted by the IRB since, for the purposes of this study, only anonymous and deidentified information were provided by the NHIS.

#### Assessment of FPG variability indices

FPG variability was calculated using FPG values of serial health examinations. Variability independent of the mean (FPG-VIM) was defined as  $100 \times \text{SD}/\text{mean}^{\beta}$ , where  $\beta$  is the regression coefficient, on the basis of the natural log (ln) of the SD over the ln of the mean. Besides FPV-VIM, we used a coefficient of variation (FPG-CV), FPG-SD, and average successive variability (FPG-ASV); defined as the average absolute difference between successive values, to evaluate the variability of FPG. The detailed description of the calculation of various FPG variability indices has been reported in previous publication.<sup>17</sup>

#### Definitions

GDM was defined as the presence of more than four times of claim of GDM (ICD-10 O24.4 and O24.9) or prescription of insulin under the ICD-code of GDM. In South Korea, the "two-step" approach with a 50g screen followed by a 100g oral glucose tolerance test for those who screen positive is mainly used for the diagnosis of GDM. Pregnancy was defined as a day before 280 days of delivery. Diabetes was defined using the ICD-10 codes E10-14 or as the level of fasting glucose of  $\geq 126 \text{ mg/}$ dL during the national health examinations. Heavy alcohol consumption was defined as drinking more than three times a week, and regular exercise was defined as moderate-intensity physical activity for more than 30 min at least 5 days a week, and high-intensity physical activity for more than 20 min at least 3 days a week. BMI was defined as body weight (kg) divided by height<sup>2</sup>  $(m^2)$ . All blood samples were drawn after more than 8 hours of fasting.



Figure 1 Flowchart of the study. GDM, gestational diabetes mellitus.

#### **Statistical analysis**

Continuous variables and categorical variables are expressed as mean±SDor as a number (percentage). To compare baseline characteristics between groups, t-test was used for continuous variables, while  $\chi^2$  test was used for categorical variables. Subsequently, the linear trend test was performed to determine the presence of a linear trend. Multivariate logistic regression analysis was used to calculate the ORs and 95% CI for the development of GDM according to the quartiles of FPG variability after adjustment for covariates including age, BMI, smoking status, heavy alcohol consumption, regular exercise, family history of diabetes, and mean FPG. The mean FPG level was defined as the average value of all fasting glucose measurements in the index year and within the previous 5 years. Sensitivity analysis was done after the exclusion of women with impaired FPG (100 mg/dL<FPG≤126 mg/dL) or diagnosed with polycystic ovarian syndrome (PCOS) during the health examination from January 2002 to pregnancy. Stratified analyses were conducted according to age (<35 years or  $\geq$ 35 years), BMI (<25 kg/m<sup>2</sup> or  $\geq$ 25 kg/m<sup>2</sup>), smoking status, heavy alcohol consumption, exercise, and family history of diabetes using the likelihood ratio test. Statistical analyses were performed using SAS V.9.4 (SAS Institute, Cary, North Carolina, USA), and p<0.05 were considered significant.

#### RESULTS

#### **Baseline characteristics**

Among the 164053 primipara with normal fasting glucose level, GDM developed in 6627 (4.04%) women who received insulin treatment (0.50%) and who did not

receive insulin treatment (3.54%) (figure 1). Women with GDM had higher age, BMI, body weight, systolic BP, diastolic BP, TC, FPG, mean FPG, FPG variability, prevalence of current smoking, and family history of diabetes than those without GDM (online supplementary table 1). Table 1 shows the baseline characteristics of women according to quartiles of FPG-VIM. BMI, systolic BP, diastolic BP, prevalence of heavy drinking, current smoker, and family history of diabetes were increased according to FPG quartile increment (p for linear trend <0.001).

#### **Risk of GDM according to the quartiles of FPG-VIM**

Table 2 shows the risk of developing GDM according to quartiles of FPG-VIM. The risk of GDM showed a stepwise increase according to the quartiles of FPG-VIM (p for linear trend <0.001). The ORs and 95% CI of GDM total were 1.10 (1.02 to 1.18), 1.13 (1.05 to 1.27), and 1.22 (1.14 to 1.31) in the second, third, and fourth quartile FPG-VIM group, respectively, compared with those in the first FPG-VIM group after adjusting for age, BMI, smoking status, alcohol consumption, exercise, family history of diabetes, and mean FPG. In particular, the risk for GDM requiring insulin therapy increased (almost 48%) in women in the highest quartile of FPG-VIM compared with those in the lowest quartile of FPG-VIM. Moreover, there was a modest increase (19%) in the risk for GDM without requiring insulin therapy in the highest FPG-VIM group. Similar OR values and trends were observed in the other indices of FPG variability assessed by CV, SD, and ASV (online supplementary table 2).

Total

|                                  | Iotai         |                |               |               |         |             |
|----------------------------------|---------------|----------------|---------------|---------------|---------|-------------|
|                                  | Q1            | Q2             | Q3            | Q4            |         |             |
| N                                | 41015         | 41011          | 41012         | 41015         | P value | P for trend |
| Age (years)                      | 30.82±3.14    | 30.77±3.10     | 30.73±3.08    | 30.66±3.08    | <0.001  | <0.001      |
| Body mass index (kg/m²)          | 20.55±2.55    | 20.62±2.60     | 20.66±2.64    | 20.71±2.69    | <0.001  | <0.001      |
| Body weight (kg)                 | 53.68±7.24    | 53.83±7.38     | 53.90±7.52    | 53.88±7.62    | <0.001  | <0.001      |
| Systolic blood pressure (mm Hg)  | 109.18±10.48  | 109.52±10.56   | 109.71±10.47  | 110.22±10.54  | <0.001  | <0.001      |
| Diastolic blood pressure (mm Hg) | 68.67±7.88    | 68.96±7.96     | 69.11±7.89    | 69.48±7.95    | <0.001  | <0.001      |
| Total cholesterol (mg/dL)        | 176.83±28.03  | 176.94±28.47   | 176.96±28.70  | 176.72±29.05  | 0.625   | 0.631       |
| Fasting glucose (mg/dL)          | 86.05±6.05    | 85.03±6.64     | 85.77±7.41    | 84.62±9.20    | <0.001  | <0.001      |
| Mean FPG (mg/dL)                 | 85.96±2.68    | 85.87±5.59     | 85.67±5.59    | 85.67±7.10    | <0.001  | <0.001      |
| FPG variability (VIM)            | 2.85±0.96     | 5.25±0.59      | 7.48±0.74     | 12.32±4.05    | <0.001  | <0.001      |
| Alcohol consumption (%)          | 1032 (2.52)   | 1127 (2.75)    | 1266 (3.09)   | 1334 (3.25)   | <0.001  | <0.001      |
| Current smoker (%)               | 2074 (5.06)   | 2244 (5.47)    | 2344 (5.72)   | 2491 (6.07)   | <0.001  | <0.001      |
| Regular exercise (%)             | 744 (1.81)    | 749 (1.83)     | 803 (1.96)    | 829 (2.02)    | 0.080   | 0.013       |
| Family history of diabetes (%)   | 30607 (74.62) | 31 799 (77.54) | 32698 (79.73) | 33118 (80.75) | <0.001  | <0.001      |

Data are expressed as the means±SD or n (%).

FPG, fasting plasma glucose; VIM, variability independent of the mean.

#### **Risk of GDM according to the deciles of FPG-VIM**

When the total participants were divided into subgroups according to deciles of FPG-VIM, there were significant linear associations between deciles of FPG-VIM and the OR of GDM (p for linear trend < 0.001) (figure 2 or online supplementary table 3). In the fully adjusted model, the

| Table 2 ORs and 95% CIs for the risk of GDM according to quartiles of FPG-VIM |            |                     |                     |                     |                     |                     |
|---|------------|---------------------|---------------------|---------------------|---------------------|---------------------|
|   | Events (n) | Unadjusted          | Model 1             | Model 2             | Model 3             | Model 4             |
| GDM total   |            |                     |                     |                     |                     |                     |
| Q1 (n=41015)  | 1484       | 1                   | 1                   | 1                   | 1                   | 1                   |
| Q2 (n=41011)  | 1621       | 1.10 (1.02 to 1.18) | 1.10 (1.03 to 1.19) | 1.10 (1.02 to 1.18) | 1.10 (1.02 to 1.18) | 1.10 (1.02 to 1.18) |
| Q3 (n=41012)  | 1656       | 1.12 (1.04 to 1.20) | 1.14 (1.06 to 1.22) | 1.12 (1.04 to 1.21) | 1.12 (1.04 to 1.20) | 1.13 (1.05 to 1.27) |
| Q4 (n=41015)  | 1866       | 1.27 (1.18 to 1.36) | 1.30 (1.21 to 1.39) | 1.27 (1.19 to 1.37) | 1.27 (1.18 to 1.36) | 1.22 (1.14 to 1.31) |
| P value   |            | <0.001              | <0.001              | <0.001              | <0.001              | <0.001              |
| P for linear trend  |            | <0.001              | <0.001              | <0.001              | <0.001              | <0.001              |
| GDM without insulin   |            |                     |                     |                     |                     |                     |
| Q1 (n=40854)  | 1323       | 1                   | 1                   | 1                   | 1                   | 1                   |
| Q2 (n=40826)  | 1436       | 1.09 (1.01 to 1.18) | 1.10 (1.02 to 1.18) | 1.09 (1.01 to 1.18) | 1.09 (1.01 to 1.18) | 1.09 (1.01 to 1.18) |
| Q3 (n=40815)  | 1459       | 1.11 (1.03 to 1.20) | 1.12 (1.04 to 1.21) | 1.11 (1.03 to 1.20) | 1.11 (1.03 to 1.20) | 1.12 (1.04 to 1.21) |
| Q4 (n=40739)  | 1590       | 1.21 (1.13 to 1.31) | 1.24 (1.15 to 1.33) | 1.22 (1.13 to 1.32) | 1.22 (1.13 to 1.32) | 1.19 (1.10 to 1.28) |
| P value   |            | <0.001              | <0.001              | <0.001              | <0.001              | <0.001              |
| P for linear trend  |            | <0.001              | <0.001              | <0.001              | <0.001              | <0.001              |
| GDM with insulin  |            |                     |                     |                     |                     |                     |
| Q1 (n=39692)  | 161        | 1                   | 1                   | 1                   | 1                   | 1                   |
| Q2 (n=39575)  | 185        | 1.15 (0.93 to 1.43) | 1.17 (0.94 to 1.44) | 1.14 (0.92 to 1.41) | 1.13 (0.91 to 1.39) | 1.13 (0.92 to 1.40) |
| Q3 (n=39553)  | 197        | 1.23 (1.00 to 1.51) | 1.25 (1.02 to 1.54) | 1.21 (0.98 to 1.49) | 1.18 (0.96 to 1.46) | 1.20 (0.98 to 1.49) |
| Q4 (n=39425)  | 276        | 1.73 (1.43 to 2.10) | 1.78 (1.47 to 2.17) | 1.70 (1.39 to 2.06) | 1.66 (1.36 to 2.01) | 1.48 (1.21 to 1.80) |
| P value   |            | <0.001              | <0.001              | <0.001              | <0.001              | <0.001              |
| P for linear trend  |            | <0.001              | <0.001              | <0.001              | <0.001              | <0.001              |

Model 1: adjusted for age.

Model 2: adjusted for model 1 plus body mass index.

Model 3: adjusted for model 2 plus smoking status, alcohol consumption, regular exercise, family history of diabetes. Model 4: adjusted for model 3 plus mean fasting plasma glucose.

FPG, fasting plasma glucose; GDM, gestational diabetes mellitus; VIM, variability independent of the mean.



Figure 2 ORs (95% CI) of newly diagnosed GDM by deciles of FPG-VIM. (A) GDM total, (B) GDM without insulin, (C) GDM with insulin. FPG, fasting plasma glucose; GDM, gestational diabetes mellitus; VIM, variability independent of the mean.

ORs for GDM total, GDM without requiring insulin treatment, and GDM requiring insulin treatment were 1.30, 1.25, and 1.58 in women in the highest decile of FPG-VIM compared with those in women in the lowest decile of FPG-VIM, respectively (figure 2 or online supplementary table 3).

#### **Sensitivity analysis**

The association between FPG-VIM and GDM was not attenuated after excluding women with FPG 100–126 mg/dL at least once during the health examinations from January 2002 to pregnancy (online supplementary table 4) or previous history of PCOS after adjusting for confounding factors including mean FPG level (online supplementary table 5).

factors of GDM including age, BMI, smoking, heavy drinking, regular exercise, and family history of diabetes (figure 3). When we compared the group in the highest quartile of FPG-VIM with those in the lowest quartile of FPG-VIM, the effect of the FPG-VIM was more obvious in the obese group than normal-weight group (p for interaction <0.001). In the aspect of age, the trend of more increased OR was noted in women aged ≥35 years than <35 years (p for interaction=0.055). By contrast, no significant interaction was observed among smoking status, heavy alcohol consumption, exercise habit, and family history of diabetes.

#### DISCUSSION

Subgroup analysis

We performed stratified analysis and interaction testing to assess the potential effect modification of the risk Our study has a unique opportunity to reveal the impact of FPG variability in the prepregnancy period on the risk of GDM in the primipara with baseline normal fasting glucose level. Women in the highest quartile of FPG-VIM

| Subgroup                   |                         | OR (95% CI)      | P for interaction |
|----------------------------|-------------------------|------------------|-------------------|
| Age                        |                         |                  |                   |
| <35                        | ⊢∎1                     | 1.12 (1.05-1.19) | 0.055             |
| ≥35                        | <b>⊢</b>                | 1.24 (1.09-1.31) |                   |
| BMI                        |                         |                  |                   |
| <25                        | <b>├-■</b> 1            | 1.11 (1.05-1.19) | < 0.001           |
| ≥25                        | <b>⊢</b>                | 1.33 (1.13-1.42) |                   |
| Current smoker             |                         |                  |                   |
| No                         | ┝╼╋╾┥                   | 1.13 (1.07-1.18) | 0.101             |
| Yes                        | <b>⊢</b> I              | 1.35 (1.02-1.55) |                   |
| Heavy alcohol consumption  |                         |                  |                   |
| No                         | <b>├-■</b>              | 1.13 (1.07-1.20) | 0.336             |
| Yes                        | <b>⊢</b>                | 1.30 (1.03-1.80) |                   |
| Regular exercise           |                         |                  |                   |
| No                         | ┝╼╋╾┥                   | 1.14 (1.08-1.64) | 0.916             |
| Yes                        | < ∎                     | 1.10 (0.74-1.63) |                   |
| Family history of diabetes |                         |                  |                   |
| No                         | k∎                      | 1.13 (0.99-1.30) | 0.530             |
| Yes                        | <b>⊢</b> ∎−-1           | 1.14 (1.07-1.21) |                   |
|                            |                         | 1                |                   |
|                            | 0.8 1.0 1.2 1.4 1.6 1.8 | 2.0              |                   |

Figure 3 Subgroup analysis for the risk of GDM in the highest quartile versus the rest three quartiles of FPG-VIM. BMI, body mass index; FPG, fasting plasma glucose; GDM, gestational diabetes mellitus; VIM, variability independent of the mean.

had an 22% increased risk for GDM even after adjusting for risk factors including mean FPG, and these associations were consistent in all indices of FPG variability such as FPG-CV, FPG-SD, and FPG-ASV. In particular, there was a 48% increase in the risk for GDM requiring insulin treatment in women in the highest quartile of FPG-VIM. The association between high FPG variability and the risk of GDM was intensified in the obese and aged more than 35 years women.

The prevalence of pre-diabetes and T2DM is more than 30% in women of reproductive-age.<sup>18</sup> However, healthy young women have less chances of being diagnosed with diabetes or impaired glucose tolerance. Although the physiologic process of GDM is complicated,<sup>19</sup> the main pathology of GDM is an aggravation of pre-existing insulin resistance which is not compensated by insulin secretion.<sup>20</sup> GDM causes various complications both in the mother and fetus, and the incidence of these complications can be reduced with treatment; hence, timely diagnosis is important. However, GDM is diagnosed at 24-28 weeks of gestational age using glucose loading test, which allows a very limited time to reverse the adverse outcome of hyperglycemia. A simple and cost-effective strategy for the early risk stratification of GDM is required.

Long-term glycemic variability using visit-to-visit FPG or HbA1C is a marker of glycemic homeostasis.<sup>9</sup> Glycemic variability has been widely studied in patients with T2DM. Increased glycemic variability correlates with diabetic polyneuropathy, retinopathy, and cardiovascular autonomic neuropathy in patients with diabetes.<sup>12 13 21</sup> A metaanalysis including eight studies reported that increased FPG variability resulted in increased all-cause mortality in diabetes (HRs: 1.28, 95% CI 1.12 to 1.46).<sup>13</sup> Recently, the impact of glycemic variability in the general population was also highlighted. Akirov *et al*<sup>22</sup> reported that increased glycemic variability is associated with longer hospitalization and increased short-term and long-term mortality irrespective of underlying diabetes. Furthermore, Oka et  $al^{23}$  demonstrated that higher glycemic variability was associated with more severe myocardial damage after percutaneous coronary intervention in non-diabetic patients with ST-segment elevation myocardial infarction. In the CARDIA study, adults aged 18-30 years without diabetes were followed up to 30 years, and elevated FPG variability was associated with the development of T2DM, CVD, and all-cause mortality.<sup>15</sup> Our study is the first to demonstrate that increased FPG variability in prepregnancy was associated with the risk of GDM in the primipara with normal fasting glucose level even after adjusting other risk factors. Although FPG variability is affected by the mean glucose level,<sup>24</sup> the association between FPG variability and GDM was consistent even after adjusting the mean FPG. Interestingly, the association of FPG variability was strengthened in insulin-treated GDM rather than in those without insulin treatment. Insulin treatment during pregnancy is the strongest risk factor for the development of T2DM;<sup>25</sup> thus, prepregnancy FPG

variability is an important risk factor for GDM and also for future glycemic derangement.

The possible pathophysiologic mechanisms connecting FPG variability and GDM could be oxidative stress and β-cell dysfunction. A previous study showed that glycemic variability measured by continuous glucose monitoring is associated with oxidative stress.<sup>26</sup> Monnier et al<sup>10</sup> demonstrated that a marker of oxidative stress was significantly elevated in patients with high glycemic variability than in those with sustained hyperglycemia. Increased glucose excursion promotes proinflammatory cytokines by oxidative stress and contributes to the development of diabetes.<sup>27</sup> In addition, intermittent hyperglycemia could induce mitochondrial superoxide overproduction and reduce adiponectin secretion.<sup>28</sup> Lower adiponectin has been a well-known marker of reduced glucose sensitivity.<sup>29</sup> Previous study demonstrated that lower adiponectin levels before pregnancy are associated with GDM.<sup>30</sup> Besides oxidative stress, insulin secretory  $\beta$ -cell function is important in the pathology of GDM. Glycemic variability causes apoptosis of pancreatic  $\beta$ -cell,<sup>31</sup> which may result in the deterioration of glycemic control and subsequent progression of diabetes. At 24-28 weeks gestation, glycemic variability correlates with impaired early-phase insulin secretion, which is higher in hyperglycemic women than in normoglycemic women.<sup>32</sup> On the contrary, glycemic instability itself might be a marker of an  $\beta$ -cell dysfunction in at-risk individuals prior to the onset of diabetes. Women with normal glucose tolerance with a history of GDM showed elevated glycemic variability with low islet  $\beta$ -cell function index.<sup>33</sup> A recent study reported that Koreans had smaller pancreatic volume and lower  $\beta$ -cell function than age and BMI matched Caucasians.<sup>34</sup> East Asians have limited capacity for insulin secretion.<sup>35</sup> Hence, glycemic variability, which suggests impaired insulin secretion, can be a more useful indicator for GDM in Asian women than in women belonging to another ethnic group. Furthermore, in our study, the association between high FPG variability and the risk of GDM was strengthened in the obese women and those aged more than 35 years, which are the well-known risk factors for GDM. Therefore, the early risk stratification of GDM using prepregnancy FPG variability might be clinically valuable, especially among obese, older, and Asian women.

Our study has several limitations. First, this study is a retrospective case-control study; we could not find any causal relationship. Second, GDM women were defined as those classified using four or more GDM codes or have taken have taken insulin during pregnancy because the nationwide database had no data on HbA1C value and oral glucose tolerance test results. However, the prevalence of GDM according to our definition was 4.2%, which correspond to that reported in previous study.<sup>36</sup> Finally, we did not consider dietary habits. Despite these limitations, our study has some strengths. It is the first to reveal the association between prepregnancy FPG variability and risk of GDM using large population-based

data. In addition, to overcome the effect of the high mean glucose level in the risk of GDM, we adjusted the mean FPG level in full adjustment models.

Taken together, high glycemic variability was related to an increased risk of GDM, and the relation remained significant after the adjustment for other risk factors. GDM affects the mother and their offspring during the perinatal period and also the life-long period; however, there are no definite ways to regress GDM after diagnosis. Therefore, monitoring the longitudinal pattern of FPG level in women of reproductive age may be clinically important for the early detection of individuals with high risk for GDM. Further studies to examine the predictive power of FPG variability for the incidence of GDM should be conducted to validate these results.

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#### **ORCID iDs**

Kyung Mook Choi http://orcid.org/0000-0001-6175-0225 Hye Jin Yoo http://orcid.org/0000-0003-0600-0266

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