Influence of dipeptidyl peptidase-4 inhibitors on glycemic variability in patients with type 2 diabetes: A meta-analysis of randomized controlled trials

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Objective: The influence of dipeptidyl peptidase-4 (DPP4) inhibitors on glycemic variability compared to other oral antidiabetic drugs (OADs), measured based on the mean amplitude of glycemic excursions (MAGE), has not been comprehensively analyzed. The aim of the study was to perform a meta-analysis to compare the effects of DPP4 inhibitors on MAGE with other OADs in type 2 diabetes mellitus (T2DM) patients without concurrent insulin treatments.

Methods: The Medline (PubMed), Embase (Ovid), and CENTER (Cochrane Library) databases were searched for relevant randomized controlled trials (RCTs). Study characteristics and outcome data were independently extracted by two authors. A random-effect model was used to combine the results.

Results: Fourteen studies with 855 patients were included. Compared to other OADs, DPP4 inhibitors significantly reduced MAGE (mean difference [MD]: -0.69 mmol/L, 95% confidence interval [CI]: -0.95 to -0.43, P<0.001) with mild heterogeneity (I² = 28%). Predefined subgroup analyses suggested that DPP4 inhibitors were more effective in reducing MAGE compared to insulin secretagogues (MD: -0.92 mmol/L, P<0.001) and non-secretagogues (MD: -0.43 mmol/L, P=0.02), as well as compared to sulfonylureas (MD: -0.91 mmol/L, P<0.001) and sodium glucose cotransporter 2 inhibitors (MD: -0.67 mmol/L, P=0.03).

Conclusions: DPP4 inhibitors may significantly reduce glycemic variability compared to other oral anti-diabetic drugs, as evidenced by MAGE in T2DM patients with no concurrent insulin treatment.
**Introduction**

Vascular complications are considered the main determinants for increased disability and mortality in patients with type 2 diabetes mellitus (T2DM) (1, 2). The major cause of vascular complication in T2DM patients is poor glycemic control and long-lasting hyperglycemia (3, 4). Interestingly, subsequent evidence showed that glycemic variability, which refers to multiple fluctuations of glycemia that occur throughout the day or over longer periods of time (5), may be another key factor that may accelerate the development of diabetic complications besides persistent hyperglycemia (6, 7). Indeed, epidemiological studies have suggested that increased glycemic variability is independently associated with the development of retinopathy, nephropathy, and possibly cardiovascular events and mortality in patients with T2DM (8–10). Experimental studies also confirmed that increased glycemic variability is associated with overactive oxidative stress and inflammatory responses (11, 12), which have been implicated in vascular complications in T2DM. Therefore, glycemic variability has been established as a novel treatment target for patients with T2DM.

Among the frequently used oral antidiabetic drugs (OADs), dipeptidyl-peptidase 4 (DPP4) inhibitors have shown promising therapeutic efficacy for T2DM patients on the basis of numerous benefits besides hypoglycemic efficacy such as the preservation of islet β-cell function (13), improvement of glycemic durability (14), and suppression of abnormal hyperglucagonaemia (15). Moreover, since increased glycemic variability is correlated with poor islet β-cell function (16), a beneficial effect of DPP4 inhibitors on glycemic variability in T2DM patients has also been suggested in view of their beneficial effects on β-cell function (17). More importantly, the pharmacological mechanisms of DPP4 inhibitors involve enhancing incretin preservation, such as glucagon-like peptide-1 (GLP-1) and glucose-dependent insulinotropic polypeptide (GIP), thereby exerting their hypoglycemic effects via glucose-dependent insulin secretion and inhibiting glucagon production without increasing the risk of hypoglycemia (18, 19).

Conventionally, glycemic variability is evaluated based on the mean amplitude of glycemic excursions (MAGEs) via continuous glucose monitoring (CGM) or flash glucose monitoring (FGM) (20) and is also considered a reliable parameter for evaluating short-term within-day glycemic variability. Although some small-scale randomized controlled trials (RCTs) have been performed to compare the influences of DPP4 inhibitors and other OADs on MAGE in T2DM patients (21–34), the results of these RCTs were not consistent and a consensus on the efficacy of DPP4 inhibitors on glycemic variability compared to other OADs has not been fully determined. A previous meta-analysis (35) including only seven RCTs before 2018 showed that DPP4 inhibitors may be more effective than other OADs in reducing MAGE. However, some recently published eligible studies (30–34) have not been included in a meta-analysis. Accordingly, in this study, we aimed to perform an updated meta-analysis to summarize the current knowledge regarding the influence of DPP4 inhibitors on glycemic variability in comparison with other OADs. With more studies included, we also explored the potential influences of the study characteristics on the outcomes.

**Methods**

The Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) statement (36) and the Cochrane Handbook guidelines (37) were followed during the design and implementation of the study.

**Data sources and searches**

Medline (PubMed), Embase (Ovid), and CENTER (Cochrane Library) databases were searched for relevant studies with the following keywords: (1) “DPP4” OR “DPP-4” OR “dipeptidyl peptidase-4 inhibitors” OR “sitagliptin” OR
“vildagliptin” OR “linagliptin” OR “saxagliptin” OR “alogliptin” OR “dutagliptin” OR “aemgliptin” OR “anagliptin” OR “teneligliptin” OR “trelagliptin” OR “omarigliptin” OR “gemiigliptin” OR “egvolgliptin”; (2) “continuous glucose monitoring” OR “glycemic variability” OR “glyceamic variability” OR “glycose variability” OR “glucose fluctuation” OR “glycemic fluctuation” OR “mean amplitude of glycemic excursion” OR “MAGE” OR “standard deviation” OR “SD” OR “SDBG” OR “largest amplitude of glycemic excursion” OR “LAGE” OR “Coefficient of variation” OR “CV”; and (3) “random” OR “randomised” OR “randomised” OR “randomly”. Only clinical studies were considered. The references of related reviews and original articles were also searched as a complement. The final database search was conducted on July 23, 2021.

Study selection

Studies that fulfilled the following criteria were included: (1) full-length articles published in English; (2) designed as parallel-group or crossover RCTs; (3) included adult patients with T2DM who were not treated with insulin; (4) patients were allocated to a treatment group with DPP4 inhibitors or a control group with other OADs; and (5) reported the between-group difference of changes in MAGE using CGM or FGM from the baseline for patients in each arm of the study. Studies with drug-naive patients or T2DM patients on background OAD therapy were included. Studies including T2DM patients with concurrent insulin therapy were excluded because the influence of insulin on blood glucose fluctuations is related to various factors such as dosage and categories, which may conceal the effect of combined OADs on blood glucose fluctuations. We did not consider studies including patients treated with single-dose/single-day DPP4 inhibitors because we did not plan to evaluate the acute effect of DPP4 inhibitors on glycemic variability. In addition, non-randomized studies, studies with non-T2DM patients, and those without MAGE measurement using CGM or FGM were excluded.

Data extraction and quality assessment

Database search, data extraction, and quality evaluation were conducted by two independent authors. Disagreements were resolved by discussion with the corresponding author. We extracted data regarding the study information (first author, publication year, and study country), study design (blind or open-label, crossover or parallel group), patient information (number of patients, mean age, sex, baseline HbA1c, and T2DM duration), details of background antidiabetic treatments, drugs and doses of DPP4 inhibitors and control OADs, treatment durations, and methods used for MAGE measurement. Quality evaluation was achieved using Cochrane’s Risk of Bias Tool (37) according to the following aspects: (1) random sequence generation; (2) allocation concealment; (3) blinding of participants and personnel; (4) blinding of outcome assessors; (5) incomplete outcome data; (6) selective outcome reporting; and (7) other potential biases.

Data synthesis and analysis

The influences of DPP4 inhibitors on MAGE compared to controls in T2DM patients were presented as the mean difference (MD) and 95% confidence interval (CI). We used Cochrane’s Q test to detect the heterogeneity (38). The I² statistic was also calculated, and an I² > 50% reflected significant heterogeneity (37). Pooled analyses were calculated using a random-effect model because this method incorporates the influence of potential heterogeneity and provides a more generalized result (37). Sensitivity analysis by excluding one study at a time was used to evaluate the influence of each study on the pooled results of the meta-analysis (37). Predefined subgroup analyses were performed to evaluate the potential influences of study characteristics on the outcome. Particularly, subgroup analysis was performed to explore the relative efficacy of DPP4 inhibitors on MAGE as compared with insulin secretagogues (sulphonylureas [SUs] and glinides) and non-secretagogues, as well as individual classes of OADs. Publication bias was evaluated by visual inspection of the funnel plots and Egger’s regression asymmetry test results (39). P values < 0.05 were considered statistically significant. The RevMan (Version 5.1; Cochrane, Oxford, UK) software was applied for statistical analyses.

Results

Search results

The process of database search and study identification is shown in Figure 1. Briefly, 2,705 articles were obtained through the database search, and 1,905 were retrieved after the exclusion of duplicate records. Among them, 1,868 articles were subsequently excluded based on the titles and abstracts primarily because these studies were irrelevant to the aim of the meta-analysis. Of the 37 articles that underwent full-text review, 23 were further excluded for the reasons presented in Figure 1. Finally, 14 RCTs (21–34) were included.

Study characteristics and data quality

Table 1 shows the characteristics of the included studies. Overall, 14 RCTs comparing the effects of DPP4 inhibitors and
other OADs on MAGE in T2DM patients were included (21–34). Since one study included two interventional groups with gemigliptin and sitagliptin, the two datasets were included in the meta-analysis independently. The studies included in the meta-analysis were published between 2013 and 2020. Three of them were crossover studies (21, 23, 32), while the others were parallel-group RCTs (22, 24–31, 33, 34). Various DPP4 inhibitors were used in these studies, such as vildagliptin, sitagliptin, linagliptin, and gemigliptin, and the control OADs included SUs, glinides, voglibose, pioglitazone, metformin, and dapagliptin. The treatment durations varied between 5 and 168 days, and MAGE was measured using CGM or FGM for 24, 48, and 72 hours. The detailed quality evaluation for these studies is shown in Table 2, suggesting moderate study quality.

**Comparisons between DPP4 inhibitors and other OADs on MAGE**

Mild significant heterogeneity was detected among the included studies (P for Cochrane’s Q test = 0.15, I² = 28%). Pooled results showed that DPP4 inhibitors significantly reduced MAGE compared to other OADs in T2DM patients with no concurrent insulin treatment (MD: -0.69 mmol/L, 95% CI: -1.21 to -0.14, P < 0.001; I² = 0%) and non-secretagogues (MD: -0.43 mmol/L, 95% CI: -0.79 to -0.07, P = 0.02; Figure 2B), as well as compared to sulfonylureas (MD: -0.91 mmol/L, 95% CI: -1.22 to -0.60, P < 0.001) and sodium glucose cotransporter 2 inhibitors ([SGLT-2] inhibitors, MD: -0.67 mmol/L, 95% CI: -1.27 to -0.08, P = 0.03, whereas the difference of MAGE between patients treated with DPP4 inhibitors and with glinides was not statistically significant (MD: -0.74 mmol/L, 95% CI: -2.11 to 0.62, P = 0.29, Figure 2C). In addition, subgroup analysis suggested that the reduction of MAGE was more remarkable in patients with T2DM duration ≤5 years as compared to those with disease duration > 5 years, and in studies with 72 hours of glucose monitoring for MAGE measurement as compared to those with 24 or 48 hours (P for subgroup difference = 0.004 and 0.002, respectively; Table 3). Other study characteristics did not seem to significantly affect the results, such as the study design, patient number, age and sex of the patients, HbA1c at baseline, background therapy, treatment duration, or quality scores of the included RCTs (Table 3). Subgroup analyses comparing DPP4 inhibitors to metformin or voglibose were not performed because only one dataset available for these comparisons.

**Publication bias**

The funnel plots for the meta-analyses comparing the effects of DPP4 inhibitors and other OADs on MAGE were
### TABLE 1 Characteristics of the included RCTs.

| Study | Country | Design | Patient number | Mean age | Male Baseline HbA1c | T2DM duration years | Background treatment | Intervention | Control | Treatment duration | MAGE measuring |
|-------|---------|--------|----------------|----------|---------------------|---------------------|----------------------|--------------|---------|-------------------|----------------|
| (21)  | USA     | R, OL, CO | 24 | 58.3 | 79.2 | 7.6 | 7 | Metformin | Vildagliptin 50mg Bid | Glimepiride 2mg Qd | 5 | 24h-CGM |
| (22)  | Korea R | DB | 33 | 57.6 | 57.6 | 7.2 | 5.4 | Metformin | Sitagliptin 100mg Qd | Glimepiride 2mg Qd | 28 | 72h-CGM |
| (23)  | Japan R | CO | 11 | 58 | NR | 7.3 | 8 | Drug naive | Sitagliptin 50mg Qd | Miglitol 10mg Qd | 28 | 24h-CGM |
| (24)  | Japan R | OL | 29 | 64.4 | 79.3 | 7 | NR | Acarbose | Sitagliptin 50mg Qd | Miglitol 5-10mg Tid | 28 | 72h-CGM |
| (25)  | China R | | 41 | 68.8 | 56.1 | 7.2 | 0 | Metformin | Sitagliptin 100mg Qd | Glimepiride 1-4mg Qd | 168 | 72h-CGM |
| (29)  | Japan R | SB | 99 | 61.4 | 40.4 | 7.4 | NR | Drug naive | Linagliptin 5mg Qd | Voglibose 0.2mg Tid | 84 | 24h-CGM |
| (28)  | Korea R | SB | 31 | 49.4 | 71 | 9.6 | 0.8 | Drug naive | Gemigliptin 50 mg Qd | Glimepiride 2mg Qd | 84 | 72h-CGM |
| (28)  | Korea R | SB | 27 | 50.5 | 75 | 9.3 | 1.6 | Drug naive | Sitagliptin 100 mg Qd | Glimepiride 2mg Qd | 84 | 72h-CGM |
| (22)  | Korea R | OL | 25 | 56.4 | 52 | 7.3 | NR | Metformin | Vildagliptin 50mg Bid | Glimepiride 2mg Qd | 112 | 72h-CGM |
| (22)  | Korea R | OL | 34 | 36 | 58.8 | 7.6 | 6.1 | Metformin | Vildagliptin 50mg Bid | Glimepiride 1-2mg Qd | 84 | 48h-CGM |
| (31)  | Brazil R | SB | 37 | 61.9 | 0 | 7.3 | NR | Metformin | Vildagliptin 50mg Bid | Gliclazide 2.5 mg Qd | 14 | 24h-CGM |
| (32)  | Japan R | CO | 51 | 51.9 | 72.7 | 7.6 | 4.9 | Metformin (750mg/d) | Linagliptin 5mg Qd | Metformin 500mg Tid | 28 | 24h-CGM |
| (33)  | Japan R | SB | 331 | 58.1 | 60.1 | 7.8 | 5.8 | Drug naive or metformin | Sitagliptin 50-100mg Qd | Dapagliflozin 5-10mg Qd | 84 | 72h-CGM |
| (34)  | Korea R | SB | 70 | 65.8 | 65.7 | 7.9 | 2.8 | Drug naive or metformin | Gemigliptin 50 mg Qd | Dapagliflozin 10mg Qd | 84 | 72h-CGM |

RCTs, randomized controlled trials; HbA1c, glycated hemoglobin; T2DM, type 2 diabetes mellitus; MAGE, mean amplitude of glycemic excursions; R, randomized; DB, double blind; SB, single blind; OL, open-label; CO, crossover; Qd, once daily; Bid, twice daily; Tid, three times per day; CGM, continuous glucose monitoring; FGM, flash glucose monitoring.

### TABLE 2 Details of study quality evaluation via the Cochrane’s Risk of Bias Tool.

| Study | Random sequence generation | Allocation concealment | Blinding of participants | Blinding of outcome assessment | Incomplete outcome data addressed | Selective reporting | Other sources of bias |
|-------|----------------------------|------------------------|--------------------------|-------------------------------|----------------------------------|---------------------|----------------------|
| (21)  | Unclear                   | Unclear                | High                     | High                          | Low                              | Low                 | Low                  |
| (22)  | Unclear                   | Unclear                | Low                      | Low                           | Low                              | Low                 | Low                  |
| (23)  | Unclear                   | Unclear                | Unclear                  | Unclear                       | Low                              | Low                 | Low                  |
| (24)  | Unclear                   | Low                    | High                     | High                          | Low                              | Low                 | Low                  |
| (25)  | Unclear                   | Unclear                | Unclear                  | Unclear                       | Unclear                          | Low                 | Low                  |
| (29)  | Unclear                   | Unclear                | Low                      | Low                           | Low                              | Low                 | Low                  |
| (28)  | Unclear                   | Low                    | High                     | High                          | Low                              | Low                 | Low                  |
| (26)  | Unclear                   | Low                    | High                     | High                          | Low                              | Low                 | Low                  |
| (27)  | Unclear                   | High                   | Low                      | Low                           | Low                              | Low                 | Low                  |
| (30)  | Unclear                   | High                   | High                     | Low                           | Low                              | Low                 | Low                  |
| (31)  | Unclear                   | High                   | High                     | Low                           | Low                              | Low                 | Low                  |
| (32)  | Unclear                   | High                   | High                     | Low                           | Low                              | Low                 | Low                  |
| (33)  | Low                       | Unclear                | High                     | High                          | Low                              | Low                 | Low                  |
| (34)  | Unclear                   | High                   | Low                      | Low                           | Low                              | Low                 | Low                  |
FIGURE 2
Forest plots for the meta-analysis comparing the influence of DPP4 inhibitors on MAGE with other OADs in T2DM patients with no concurrent insulin therapy: (A) forest plots for the overall meta-analysis; (B) forest plots for the subgroup analysis according to the control OADs; and (C) forest plots for the subgroup analysis according to the individual OAD class of controls.
symmetrical, suggesting a low risk of publication bias (Figure 3). The results of Egger’s regression tests also suggested a low risk of publication bias (P = 0.301).

**Discussion**

In this meta-analysis, by pooling the results of 14 RCTs, we found that DPP4 inhibitors were more effective than other OADs in attenuating glycemic variability in T2DM patients without concurrent insulin treatment, as evidenced by more significantly reduced MAGE after treatment with DPP4 inhibitors. The results of sensitivity analysis by excluding one dataset at a time showed consistent results, reflecting the stability of the findings. Interestingly, the results of subgroup analyses suggested that DPP4 inhibitors were more effective in reducing MAGE compared to insulin secretagogues and non-secretagogues, as well as compared to sulfonylureas and SGLT-2 inhibitors, in patients with T2DM duration ≤ 5 years as compared to those with disease duration > 5 years, and in studies with 72 hours of glucose monitoring for MAGE measurement as compared to those with 24 or 48 hours. Taken together, this meta-analysis confirmed that DPP4 inhibitors may significantly reduce glycemic variability as evidenced by MAGE in T2DM patients with no concurrent insulin treatment.

Currently, a number of parameters have been applied to evaluate the extent of glycemic variability. Among them, time in target range (TIR) is a new vital index of glycemic control that was recently proposed to describe the percentage of time when...
glucose values fall within the target range (the common default range of 3.9–10 mmol/L) in 24 h (40). Although this index is highly recommended by the American Diabetes Association (ADA) guidelines because TIR can not only be used to assess blood glucose fluctuation but also be used as a glycemic target besides HbA1c, studies using this parameter to evaluate the effect of anti-diabetic drugs on glycemic variability are still emerging (41). MAGE was considered as a gold standard for assessing the short-term within-day glycemic variability (GV), so that a large number of high-quality studies that used MAGEs for the evaluation of glucose fluctuation have been performed (42, 43). Previous studies showed that MAGE is independent of mean glycemic level (42) and associated with higher circulating insulin antibody (44), which may provide additional information for the evaluation and management of patients with T2DM (45). A cross-sectional study including 288 Chinese T2DM patients suggested that MAGE was better associated with vascular outcomes including coronary artery disease (CAD), stroke, and chronic kidney disease (CKD) than other indices of glycemic variability (46). Besides, previous studies showed that high MAGE was an independent predictive factor of poor prognosis in patients with CAD (9), including acute coronary syndrome (47, 48), and an early study showed that glycemic variability expressed as the MAGE was significantly associated with coronary plaque rupture in patients with acute myocardial infarction (AMI) (49). In addition, it was shown that high MAGE level was significantly associated with incidence of major adverse cardiovascular events after AMI (50). Similarly, in T2DM patients with acute ischemic stroke, MAGE was shown to be a predictor of early neurological deterioration (51). However, to the best of our knowledge, a clinically significant change margin of MAGE in patients with T2DM remains unknown, which needs more studies to investigate in the future. In addition, increased glycemic fluctuation is just one of the potential factors to the incidence of MACE in CVOTS. Nearly all CVOTs conducted with DPP4 inhibitors so far have demonstrated a neutral effect on MACE (52–56).

To the best of our knowledge, there is only one meta-analytic study that evaluated the influences of DPP4 inhibitors on glycemic variability compare to other OADs (35). This study included seven RCTs before 2018 and concluded that DPP4 inhibitors may be more effective than other OADs in reducing MAGE. The limited number of available datasets prevented the authors from performing further subgroup analyses. Moreover, the previous meta-analysis included a study with a methodological flaw in randomization (57), which significantly confounded the results of the meta-analysis. Compared to the previous meta-analysis, the strengths of the current meta-analysis include the rigorous literature search, strict inclusion and exclusion criteria, and performance of multiple sensitivity analysis to confirm the stability and robustness of the findings. We included 14 RCTs with 855 patients with T2DM, which was much larger than the previous meta-analysis (279 patients). The results suggested that DPP4 inhibitors are more effective than other OADs at attenuating glycemic variability in T2DM patients with no concurrent insulin therapy. This is of clinical importance. The available OADs for patients with T2DM have increased in recent decades. Although these OADs all have glucose-lowering efficacy, some may provide additional benefits for T2DM patients. In view of the potential role of increased glycemic variability in the pathogenesis of vascular complications in T2DM (12), agents that reduce glycemic variability may further improve the prognosis of T2DM patients.

The specific multidimensional mechanisms of DPP4 inhibitors for glycemic regulation, such as enhancing incretin preservation and inhibiting glucagon production (18, 19), may
be the key reason that DPP4 inhibitors are more effective than other OADs in reducing MAGE. Further subgroup analysis according to individual OAD categories for the control showed that DPP4 inhibitors were more effective than SUs at reducing MAGE, whereas DPP4 inhibitors were comparable to glinides at reducing MAGE. This is consistent with previous observations that SUs are associated with higher glycemic fluctuation in T2DM patients (58). Moreover, glinides are short-acting insulin secretagogues and generally administered three times daily, which is also shown to be associated with reduced glucose fluctuation as compared with SUs (59). Interestingly, we also found that DPP4 inhibitors are more effective than SGLT-2 inhibitors in reducing MAGE. Of note, two of the included studies with SGLT-2 inhibitors as controls both used dapagliflozin. A recent experimental study in high-fat, high-sucrose diet mice showed that dapagliflozin increased plasma glucagon levels during acute declines in blood glucose (60), which may lead to increased glycemic fluctuation. Since only two RCTs were included in the subgroup of SGLT-2 inhibitors, further studies are needed to validate the findings and clarify the possible mechanisms underlying the superiority of DPP4 inhibitors to SGLT-2 inhibitors on MAGE.

Subgroup analysis also suggested that the reduction of MAGE was more remarkable in patients with T2DM duration ≤ 5 years as compared to those with disease duration > 5 years. Patients with longer T2DM duration are likely to have increased glycemic fluctuation severity, which is likely correlated with impaired pancreatic beta-cell function in these patients (61). Subsequently, the severe glycemic dysregulation in patients with long T2DM duration challenges hypoglycemic treatment to minimize glycemic fluctuations. In addition, a more remarkable reduction in MAGE after DPP4 inhibitor treatment was shown in studies with glucose monitoring for 72 hours as compared to those with 24 or 48-hour monitoring. These results indicate that the improvement of glycemic variability by DPP4 inhibitors was even more remarkable with a relatively longer monitoring duration.

This study also has limitations that should be considered when the results are interpreted. Firstly, there remains no consensus regarding the optimal parameter for measuring glycemic variability in T2DM patients. The influence of DPP4 inhibitors on other glycemic variability parameters in T2DM patients, such as TIR, should be evaluated. Besides, GV was analyzed based on CGM with relatively short durations among the included studies (24–72 hours). Parameters of GV calculated on the basis of CGM with longer duration (e.g., 14 days) may more accurately reflect the severity of glucose fluctuation. In addition, this meta-analysis was based on data at the study level rather than individual patients. Therefore, the results of subgroup analysis should be interpreted with caution. Large-scale RCTs or meta-analysis based on individual patient data may be considered to confirm whether patient characteristics or concurrent medications influence the potential effect of DPP4 inhibitors on MAGE. Finally, nine of the 14 RCTs included in the meta-analysis used insulin secretagogues (mostly SUs) as controls. None of the included studies compared DPP-4 inhibitors with the oral formulation of the GLP-1 receptor agonist semaglutide on GV. Studies are warranted for further investigation and the conclusion that DPP-4 inhibitors may be more effective than other OADs in reducing GV should be interpreted with caution in this regard.

Conclusions

In conclusion, the results of the meta-analysis showed that DPP4 inhibitors may be more effective than other OADs in attenuating glycemic variability in T2DM patients receiving no concurrent insulin therapy.

Data availability statement

The original contributions presented in the study are included in the article/supplementary material. Further inquiries can be directed to the corresponding author.

Author contributions

SC, RZ, RC, YZ and MY conceived and designed research; SC and RZ collected data and conducted research; SC and MY performed or supervised analyses; SC, RZ, YMZ, SR and MY interpreted data; SC wrote the initial paper. All authors provided substantive suggestions for revision, reviewed and approved final version of the paper, and for all aspects of the work in ensuring that questions related to the accuracy.

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Conflict of interest

SC, RZ, YZ and YMZ are employees of MSD China; SR is employee of Merck Sharp & Dohme LLC, a subsidiary of Merck & Co., Inc., Rahway NJ, US. This study received funding from MSD China. The funder had the following involvement with the study: study design, data collection and analysis, and preparation of the manuscript. All authors declare no other competing interests.

The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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