The Relationship between Composite Measures of Glycemic Control and Beta-Cell Function in Type 2 Diabetes

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Abstract

Objective: To investigate the relationships between indices of the quality of glycemic control and beta-cell function and insulin sensitivity, using composite metrics derived from continuous glucose monitoring data and measures of beta cell function obtained from liquid meal tolerance tests (LMTT).

Design and Methods: This is a retrospective, observational, cross-sectional study that included 56 noninsulin treated subjects with type 2 diabetes. We used continuous glucose monitoring and LMTT data after withdrawal of oral antidiabetic drugs for computation of the composite glycemic control metrics Q-score, ADRR, and GRADE. The LMTT-based disposition index calculated from the ratio of the area-under-the-insulin-curve to the area-under-the-glucose-curve provided the estimate of beta cell function in relation to the insulin sensitivity. The Matsuda index served as correlate for the whole-body insulin sensitivity.

Results: We demonstrated inverse, curvilinear relationships between the composite metrics of glucose control and beta cell function. The LMTT-DI was the main predictor variable of the Q-score, ADRR, and GRADE with diabetes duration as the minor one (Q-score: LMTT-DI, β=-0.018 ± 0.002, p<0.001; diabetes duration, β=0.267±0.082, p =0.002, R²=0.598), whereas the Matsuda insulin sensitivity index did not contribute in these multivariate regression models.

Conclusion: The strength of association between the investigated composite metrics of glucose control and the beta cell indices enables assessment of the beta cell reserve from continuous glucose monitoring data for optimization of glycemic control before therapy escalation is considered. Our results support the view that maintaining or restoring of beta cell function is pivotal to improve diabetes treatment. The use of composite metrics simplifies clinical judgment through stratification of the quality of glycemic control into Good, Satisfactory, Fair, and Poor.

Keywords: Q-score; ADRR; GRADE; Disposition index; Insulin secretion index; Matsuda insulin sensitivity index; Type 2 diabetes

Abbreviations: ADRR: Average Daily Risk Range; CGM: Continuous Glucose Monitoring; Gluc: Glucose; GRADE: Glycemic Risk Assessment Diabetes Equation; AUC: Total Area Under the Curve AUCins/gluc ; ISI : Insulin Secretion Index; Ins: Insulin; LMTT-DI: Liquid Meal Tolerance Test-Based Disposition Index; %GRADEhyp and %GRADEhyper: Percentage of GRADE in Hypo- and Hyperglycemia; HBGI: High Blood Glucose Index; ISI: Insulin Secretion Index; LBGI: Low Blood Glucose Index; LMTT-DI: Liquid Meal Tolerance Test-based Disposition Index; Matsuda Index: Index of Insulin Resistance; Q-score: Quality Score Of Glycemic Control

Introduction

Since the more widely use of continuous glucose monitoring (CGM), researchers and clinicians became aware of insufficiencies in the use of HbA1c as the sole measure of the quality of glycemic control in diabetes. Percentile scores for hypo- and hyperglycemia or composite markers gained increasing attention in the management of the large amounts of data provided by CGM [1-3]. Proposed composite markers are, for example, ADRR [4], GRADE [5], the “glucose pentagon” [6], the hypoglycemia-A1c score [7] or the Q-score [8], combining information from the overall glycemic control, hypo- and hyperglycemia, and several types of glycemic variability. The recently introduced personal glycemic state (PGS) is a composite metric [9] similar to our Q-score.
Such measures could help assess the efficacy of therapeutic interventions and individualize diabetes treatment. Although used in a few studies as glycemic outcomes measures [10-15] the application of composite measures in research and clinical practice is as yet limited [16]. The implication of metabolic factors other than metrics of glycemia possibly reflected by or closely related to these measures could provide a better understanding and broaden their application. All the variables combined in the other composite measures are apparently more or less related to beta cell dysfunction and insulin resistance, even though, the degree of their association is not well known. We previously showed that in type 2 diabetes (T2D) glycemic variability is strictly related to postprandial beta cell function [17], and others demonstrated relationships between glycemic variability and functional variables in type 1 diabetes [18]. Thus, it is conceivable that these composite measures of the quality of short-term glycemic control may provide meaningful data for the assessment of the patients’ beta cell functional reserve and insulin sensitivity. Whether the other quality metrics can complement or even replace measures of beta cell function and insulin action in detecting the progression of the disease process or the efficacy of the therapeutic intervention, has not been investigated. We hypothesize that the composite measures do not only reflect short-term changes in glycemia but underlying functional defects of the beta cell and insulin sensitivity during the worsening of diabetes control. Addressing this issue, we used composite markers of glycemic control derived from continuous glucose monitoring and the liquid meal tolerance test-based disposition index (LMTT-DI) for evaluation of beta cell function and the Matsuda index as the measure of insulin sensitivity.

Subjects and Methods

Database

The present investigation is cross-sectional and used anonymized data obtained from patients with type 2 diabetes (T2D). For data analysis, we applied validated measures of glycemic control derived from retrospective ambulatory continuous glucose monitoring (CGM) profiles (MiniMed Solution Software, Medtronic MiniMed) and used established indices of beta cell function derived from liquid meal tolerance tests. After a 12h overnight fast, all patients had received a standardized 500 mL liquid meal consisting of 75g carbohydrate, 58g fat, and 30g protein to total 1,000 kcal. The patients had CGMs performed and the Matsuda index as the measure of insulin sensitivity.

Table 1: Clinical and laboratory data. Data are mean ± SD or median (25th-75th percentile).

| Number of Patients | 56 |
|--------------------|----|
| Age (years)        | 64.4±8.6 |
| Sex men/women      | 29/27 |
| Body mass index (kg/m²) | 29.9±3.7 |
| Waist circumference (cm) | 102.8±12.6 |
| Diabetes heredity, No (%) | 27 (40.2) |
| Known diabetes duration (years) | 5.5 (2.0-10.0) |

| Blood pressure (mmHg) |
|-----------------------|
| Systolic              | 135.0 (130.0 -150.0) |
| Diastolic             | 80.0 (80.0-85.0) |

Indices of quality of glycemic control

The best of these were the Low Blood Glucose Index (LBGI), High Blood Glucose Index (HBGI), average daily risk range (ADRR) [3], Glycemic Risk Assessment Diabetes Equation (GRADE), % GRADE hypoglycemia, % GRADE hyperglycemia [4], and the newly developed Q-score [8]. The Q-score integrates the variables mean glucose, glucose range, time in glycemia (tG <3.9 mol/L, and >8.9mmol/L, and mean of daily differences (MODD). Factor analysis allowed the selection of these variables. To ensure equivalent weight the variables were standardized by z-transformation, and the Q-score was then computed according to the formula [8], as previously described:

\[
\text{Q-Score} = 8 + \frac{\text{mean glucose} -7.8}{1.7} + \text{range} -7.5 + \frac{tG>8.9 + \text{MODD} - 1.8}{5.7} + \frac{tG>9.9}{0.9}
\]

In this formula, hypoglycemia indicates time spent below the range <3.9mmol/L (tG <3.9) and time spent above the target range >8.9mmol/L (tG >8.9) is defined as hyperglycemic.
respectively. Glycated hemoglobin (HbA1c) was determined by standard methods and used as glucose exposure measure.

**Indices of beta cell function and insulin sensitivity**

Before commencing the study procedure, blood glucose lowering medication was withdrawn and substituted with placebo for eight days. We used plasma glucose and insulin concentrations obtained during the liquid meal tolerance test (LMTT) to calculate indices of insulin secretion and insulin sensitivity [19,20]. We calculated the Insulin secretion index (total $AUC_{\text{ins/gluc}}$) as total area under the curve for plasma insulin (pmol/L x150min) from 0 to 150min divided by the total area under the curve for plasma glucose (mmol/L x150min) from 0 to 150min. The areas under the curve (AUC) were calculated using the trapezoidal method. Fasting glucose ($Gluc_0$) and fasting insulin ($Ins_0$) and the respective LMTT mean post meal glucose mean glucose and mean insulin concentrations provided the data for computation of the Matsuda insulin sensitivity index (Matsuda index) according to the formula $10,000/\sqrt{Gluc_0 \times Ins_0 \times \text{mean glucose} \times \text{mean insulin}}$. Beta cell function taking the prevailing insulin sensitivity into account was assessed by calculation of the liquid meal tolerance test-based disposition index (LMTT-DI) according to the formula: Matsuda index x total $AUC_{\text{ins/gluc}}$ [20].

**Statistical analysis**

The variables are presented either as means ± SD or as medians (25th-75th percentile). Insulin is given as pmol/L and glucose as mmol/L. For comparison of continuous variables, we used either one-way analysis of variance (ANOVA) or Kruskal-Wallis one way analysis, as appropriate. Control of multiple pairwise comparisons was performed using the Holm-Sidak and Dunn’s method, respectively. We used Spearman’s correlation analysis, best subset regression with Mallow Cp statistic, and multivariate regression models to relate beta cell function and insulin sensitivity to indices of the quality of glycemic control. Potential confounding variables were age, sex, diabetes heredity, diabetes duration, and waist circumference. A variance inflation factor of and Durbin-Watson statistic served as the criterion for co-linearity. $P<0.05$ was considered statistically significant. The statistical analysis used the Statistical Package for the Social Sciences software package (version 17.0; SPSS, Chicago, IL)

**Results**

**Table 2:** Correlation matrix presented as spearman correlation coefficients $r$ for ß-cell function and insulin resistance with quality indices of glucose control. Values significant at the $p <0.001$ level (two sided) are shown in bold font.

|               | ISI   | Matsuda index | LMTT-DI | Q-score | ADRR | LBGI | HBGI | GRADE hypo | %GRADE hyper |
|---------------|-------|---------------|---------|---------|------|------|------|------------|--------------|
| Matsuda index | -0.355| 1             |         |         |      |      |      |            |              |
| LMTT-DI       | 0.571 | -0.487        | 1       |         |      |      |      |            |              |
| Q-score       | -0.484| -0.383        | -0.789  | 1       |      |      |      |            |              |
| ADRR          | -0.474| -0.396        | -0.810  | 0.949   | 1    |      |      |            |              |
| LBGI          | 0.499 | 0.162         | -0.704  | -0.432  | -0.514| 1    |      |            |              |
| HBGI          | -0.547| -0.405        | -0.892  | 0.915   | 0.941| -0.670| 1    |            |              |
| GRADE         | -0.558| -0.397        | -0.896  | 0.905   | 0.926| -0.690| 0.99 | 1          |              |
| %GRADE hypoglycemia | 0.304 | 0.09         | 0.44    | -0.13   | -0.23| 0.715 | -0.407| -0.407 | 1            |
| %GRADE hyperglycemia | -0.512 | 0.407     | -0.869  | 0.897   | 0.921| -0.682| 0.975| 0.042 | 1            |

Table 2 shows the degree of correlation between the scores of quality of glycemia and the indices of beta cell function. On Spearman’s correlation analysis, the strength of associations of these scores with the LMTT-derived disposition index (LMTT-DI) was high; moderate with the insulin secretion index (total $AUC_{\text{ins/gluc}}$), and weak with the Matsuda insulin sensitivity index. Of note, we found a stronger correlation of LMTT-DI with the low blood glucose index LBGI than with the %GRADE hypoglycemia index.

The beta cell function indices, LMTT-DI and insulin secretion index were associated with the quality of glycemic control. Figure 1 demonstrates that LMTT-DI and insulin secretion index decreased with increasing Q-score categories and was lowest in the state of glycemia defined as poor. Even though the Matsuda insulin sensitivity index did significantly differ between good ($\leq 5.9$) and satisfactory (6.0-8.4) glycemic control, it remained rather unchanged in the higher Q-score categories. Figure 2 reveals that the relationship between the Q-score and beta cell function, as defined by the LMTT-DI, was inverse, nonlinear ($r=0.873$, $p<0.001$). A decrease in LMTT-DI below 255 (1/mmol²) resulted in a steep increase in the Q-score values. We also observed significant curvilinear associations of similar magnitude (not shown) between LMTT-DI and ADRR ($r=0.865$, $p<0.001$) and GRADE ($r=0.908$, $p<0.001$). To assess the independent effects of the LMTT-DI beta cell function and the Matsuda index on the quality of glycemic control, we chose the Q-score and applied the best subset regression analysis with Mallow Cp statistic, including possible confounders such as sex, age, diabetes heredity, diabetes duration, and waist circumference. The significant variables obtained from this analysis entered a multiple regression model, which verified the LMTT-DI and diabetes duration as predictor variables of the Q-score (LMTT-DI).
Discussion

We have shown in this retrospective study that composite metrics of the quality of glycemic control such as the Q-score, ADRR, and GRADE depend primarily on the beta cell function, as indicated by the LMTT-derived disposition index. The curvilinear relationship between the LMTT-DI and the quality indices of glycemia implies that the better the beta cell reserve, the better the glucose control (lower quality indices). As exemplified by the Q-score, the beta cell functional capacity at $< 255 (1/mmol²)$ associated with a drastic increase toward poor glycemic control (Q-score levels $\geq 12.0$). The disposition index LMTT-DI assesses the insulin secretion related to the whole-body insulin sensitivity and is an important, validated measure to quantify beta cell function [21]. It was an unexpected finding that the Matsuda insulin sensitivity index correlated weakly with the composite indices of the quality of glycemia and did not significantly contribute to their variance. Except for diabetes duration, the confounding factors sex, age, diabetes heredity, and waist circumference did not affect the quality indices of diabetes control. The beta cell function was the dominant predictor variable, explaining 52, 48, and 60% of the variance in the Q-score, ADRR, and GRADE, whereas the diabetes duration defined only 8-9%. Park et al. [22] reported weaker correlations between ADRR and HOMA-%B in patients with T2D than we found between ADRR and LMTT-DI in our diabetes cohort. The explanatory variables LMTT-DI and diabetes duration could not entirely elucidate the interindividual variability of the quality of glycemic control metrics. It appears that other factors such as carbohydrate consumption, treatment duration or modality of baseline antidiabetic therapy may partly be responsible for the residual variability. Since the study participants did not receive standardized diet, carbohydrate consumption was not accurately assessable. Although the immediate pharmacological effects of antidiabetic drugs were washed out, glucose toxic effects affecting the beta cell prevailed.

Limitations of the current investigation are the retrospective nature and the relatively small number of study participants. The strength is that we assessed the composite metrics of the quality of glycemia and the indices of beta cell function as well as insulin sensitivity during oral drug withdrawal to mostly eliminate therapeutic influences [17].

Conclusion

In summary, we demonstrate the existence of inverse, curvilinear relationships between composite metrics of glucose control and beta cell function. The strength of association between the investigated composite glucose control measures and the beta cell indices was comparable. The close relationships, described herein, enable assessment of the beta cell reserve for optimization of glycemia before therapy escalation is considered. Furthermore, our results support the view that maintaining or restoring of beta cell function is pivotal to improve the quality of diabetes control, which is simplified by the use of composite metrics through the judgment Good, Satisfactory, Fair, and Poor.
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