Early and ongoing stable glycaemic control is associated with a reduction in major adverse cardiovascular events in people with type 2 diabetes: A primary care cohort study

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Abstract

Aim: To determine whether achieving early glycaemic control, and any subsequent glycaemic variability, was associated with any change in the risk of major adverse cardiovascular events (MACE).

Materials and Methods: A retrospective cohort analysis from the Oxford-Royal College of General Practitioners Research and Surveillance Centre database—a large, English primary care network—was conducted. We followed newly diagnosed patients with type 2 diabetes, on or after 1 January 2005, aged 25 years or older at diagnosis, with HbA1c measurements at both diagnosis and after 1 year, plus five or more measurements of HbA1c thereafter. Three glycaemic bands were created: groups A (HbA1c < 58 mmol/mol [<7.5%]), B (HbA1c ≥ 58 to 75 mmol/mol [7.5%-9.0%]) and C (HbA1c ≥ 75 mmol/mol [≥9.0%]). Movement between bands was determined from diagnosis to 1 year. Additionally, for data after the first 12 months, a glycaemic variability score was calculated from the number of successive HbA1c readings differing by 0.5% or higher (≥5.5 mmol/mol). Risk of MACE from 1 year postdiagnosis was assessed using time-varying Cox proportional hazards models, which included the first-year transition and the glycaemic variability score.

Results: From 26 180 patients, there were 2300 MACE. Compared with group A->A transition over 1 year, those with C->A transition had a reduced risk of MACE (HR 0.75; 95% CI 0.60-0.94; P = .014), whereas group C->C had HR 1.21 (0.81-1.81; P = .34). Compared with the lowest glycaemic variability score, the greatest variability increased the risk of MACE (HR 1.51; 1.11-2.06; P = .0096).

Conclusion: Early control of HbA1c improved cardiovascular outcomes in type 2 diabetes, although subsequent glycaemic variability had a negative effect on an individual's risk.
Cardiovascular disease (CVD) is a major cause of death and disability in people with diabetes and CVD risk is associated with higher HbA1c. The UK Prospective Diabetes Study (UKPDS) showed that, even after subsequent deterioration in glycaemia, early tight glycaemic control was associated with a reduced number of macrovascular events. Glycaemic control achieved soon after the diagnosis of type 2 diabetes has been associated with a reduction in the risk of subsequent macrovascular events, whereas a deteriorating HbA1c pattern after diagnosis is associated with an increased risk of long-term co-morbidities and death. These findings led to the concept of a ‘legacy effect’ (i.e. metabolic memory) of early glycaemic control. The mechanisms(s) underpinning the legacy effect are unresolved but may relate to epigenetic processes such as persistence of post-translational histone methylation and changes in microRNA after normalization of glucose, or the prevention of glycation of the protein matrix in arterial walls, thereby curbing the development of atherosclerotic lesions.

Conversely, the introduction of tight glycaemic control only later in the disease course—several years after the diagnosis of type 2 diabetes—may lead to adverse cardiovascular outcomes. Greater glycaemic variability has been implicated as a pathophysiological mechanism. Fluctuation in daily glucose values, or longer term swings (visit-to-visit HbA1c), may lead to oxidative stress, promoting inflammation, endothelial dysfunction and greater macrovascular risk and death. The importance of glycaemic variability on vascular outcomes in type 2 diabetes has been challenged. It is unclear whether glycaemic variability affects future macrovascular risk once the early period of glycaemic control (relating to the legacy effect) has been accounted for. To date, few studies have distinguished between the metabolic flux needed to achieve early glycaemic control, from later glycaemic variability. We used the Royal College of General Practitioners (RCGP) Research and Surveillance Centre (RSC) database to perform a comprehensive examination of glycaemic control achieved within the first year of diagnosis and subsequent HbA1c variability, with incident CVD.

1 | INTRODUCTION

2 | MATERIALS AND METHODS

2.1 | Source population

This was a retrospective cohort analysis of adults with type 2 diabetes identified from the RCGP RSC database. This comprises primary care data from a sentinel network of general practices distributed across England. The clinical computing systems that the contributing practices use are EMIS Web, INPS Vision and TPP SystmOne. At the time of analysis, the database contained primary care records collected from 1,595,170 people registered with 164 practices (2.9% of the population of England). The database included all recorded clinical codes with associated values and dates for the population up to 31 December 2016. At the time of the study, clinical codes were recorded using the Read code 5-byte version 2 and Clinical Terms Version 3 coding hierarchies and included diagnosis codes, medication codes, investigation codes, process of care codes and laboratory data. Since 2018, this has been replaced by Systematized Nomenclature of Medicine—Clinical Terms (SNOMED-CT). The RCGP RSC has been shown to be representative of the national population with regards to demographic composition, geographical coverage and prevalence of chronic disease.

UK general practice is a registration-based system with citizens registering with a single general practitioner. Care is free and nearly all care and prescribing for type 2 diabetes is carried out in primary care. Patients with type 2 diabetes are identified through an algorithm as previously reported. The first step identifies all people with diabetes (of any type) through (i) diagnostic code (diagnosis of diabetes), (ii) clinical investigations (two or more fasted, random or glucose tolerance test values or HbA1c measurements consistent with diagnosis), or (iii) medication use (two or more prescriptions for oral diabetes medications, excluding metformin or injectable therapies). These people were then categorized by diabetes type using a clinically based seven-step algorithm, centred on, but not limited to, the duration of preceding oral glucose lowering medication use.

2.2 | Design

We constructed a time-varying Cox proportional hazards model for time to major adverse cardiovascular events (MACE). A priori we had considered that it was more appropriate to consider HbA1c fluctuation as being time-dependent (i.e. the variance of HbA1c would not be expected to be uniform throughout the follow-up period). The measurement of time was the number of years since diabetes diagnosis.

2.3 | Inclusion and exclusion

We excluded patients who did not have any HbA1c results within 3 months either before or after diagnosis and who did not have a second HbA1c result during the first year after diagnosis because we were unable to classify their levels of early HbA1c exposure. We also excluded individuals with less than five HbA1c measurements.
thereafter, as this was required for assessment of HbA1c variability. MACE within the first 12 months of diagnosis of type 2 diabetes were excluded as they were considered to not represent the effectiveness, or otherwise, of any legacy effect.

2.4 Exposure

We followed newly diagnosed patients with type 2 diabetes (first entry of diabetes indicator at any time on or after 1 January 2005 to 31 December 2016). To be included, patients had to be aged 25 years or older at diagnosis and have HbA1c measurements at both diagnosis and 1 year after (a leeway of 3 months was applied either side of diagnosis and at the 1-year follow-up time point). Furthermore, all patients had to have an additional five or more measurements of HbA1c, after the first year, to be included.

The first registered HbA1c measurement within 3 months either side of first indicator code for diabetes was used as the index value. The HbA1c measurement 12 months after the first measurement (±3 months) was used as the ‘1-year’ HbA1c. If an individual had more than one measurement of HbA1c in the 9-15 months period after the first measurement, the latest registered measurement in that time span was used as the 1-year value.

We constructed an indicator variable for glycaemic transition over the first year by banding the cohort into three groups based upon their initial HbA1c level (mmol/mol). Categories of HbA1c were chosen from reports of a J-shaped relationship of HbA1c with macrovascular risk with a broadly comparative risk for HbA1c from 7.5% to 9.0%.

The categories used were: group A (HbA1c < 58 mmol/mol [<7.5%]), group B (HbA1c ≥ 58 and < 75 mmol/mol [7.5%-9.0%]) and group C (HbA1c ≥ 75 mmol/mol [≥9.0%]).

We then also recorded the group status, per individual, at the end of the first year and categorized the glycaemic transition that had occurred. For example, a person with status A->A had an HbA1c level below 57 mmol/mol at their first measurement and at 1 year. A person with status C->A would initially have had an HbA1c level above 75 mmol/mol, but then reduced this at the end of year 1 to be lower than 57 mmol/mol.

The models also included a glycaemic variability score based upon the method reported by Forbes et al., which we modified by using time-varying covariates. In the sensitivity analysis, HbA1c was treated as a time-fixed approach, whereby the covariate is considered to be more constant through the follow-up period. In brief, to calculate a glycaemic variability score using HbA1c values, the number of times successive HbA1c readings differed by 0.5% or more (≥5.5 mmol/mol) were counted. This number was divided by the number of comparisons, then multiplied by 100. For example, if a person had a sequence of HbA1c values of 6.7%, 7.0%, 7.8%, 7.4%, 8.0% and 7.9%, the number of times that a difference of 0.5% or more was noted would be two and the score would be 40 (i.e. [100 × 2]/5). As with Forbes et al., for analysis purposes, we grouped the scores into five categories: 0-20, 21-40, 41-60, 61-80 and 81-100.

Measurements during the first 12 months were excluded from the variability score to distinguish the effect of early glucose normalization from later glucose variability.

2.5 Outcomes

The outcomes of interest were the first occurrence of MACE, defined as myocardial infarction (MI), coronary intervention, stroke and amputation/limb revascularizations. All patients were followed for outcomes until the earliest date of the following: death, deregistration from practice, practice leaving the RCGP RSC network or 31 December 2016.

2.6 Covariates

We adjusted for potentially confounding variables using the most recently recorded measurement at the time of diagnosis of type 2 diabetes. Baseline covariates in the model were age (banded into the following categories: 25-44, 45-64, 65-74, 75-84 and ≥85 years), smoking status, body mass index (BMI), total cholesterol, HDL-cholesterol and LDL-cholesterol. When there were issues with missing data, the last observation was carried forward. Co-morbidities were defined using codes for diagnosis, investigation and process of care (Table S1).

2.7 Statistical analysis

We summarized data for patient characteristics using means and standard deviations (SDs) for continuous variables and counts and percentages for categorical variables. We used the chi-squared test for bivariate statistics for categorical variables and the Mann–Whitney U test or t-test for non-parametric or parametric, continuous variables, respectively.

Analyses addressing either change in HbA1c over the first year, or HbA1c variability, were adjusted for the other. HbA1c was treated as a time-dependent variable in the model with variability based on prior data and the hazard ratio interpreted as giving an instantaneous hazard of MACE. Each person was followed until death, leaving the RCGP RSC database, for instance, by emigration (censoring), or 31 December 2016, whichever came first. This was a complete-case analysis. Missing data were handled by multiple imputation and the multiple results then combined into one inference using Rubin’s rules.

A sensitivity analysis was performed that used HbA1c as a time-fixed variable.

Estimates are presented with hazard ratios (HR) and 95% confidence intervals. A two-sided P of less than .05 was considered statistically significant for all analyses. All analyses were performed using R statistical software version 3.5.3.

2.8 Compliance with ethics guidelines

All data were pseudonymized at the point of data extraction. No clinically identifiable information was available to researchers. National Research Ethics Committee approval was obtained on 30 September.
3 | RESULTS

3.1 | The diabetes cohort

A total of 1,595,170 people were included from 164 primary care practices. Within this total population, a cohort of 90,730 (5.7%) adults were identified as having diabetes. Of these, 84,378 (93.0%) individuals were categorized as having type 2 diabetes. Within this cohort we only included individuals who had diabetes diagnosed on or after 1 January 2005, with HbA1c measurements at diagnosis and 1 year after, plus at least five other times thereafter (n = 26,180) (Table 1). Most exclusions were a result of pre-existing diabetes (prior to 2005). The characteristics of the other excluded patients (n = 4,280) are listed in Table S2. The median duration of follow-up was 1583 (IQR 987-1737) days.
At the time of data extraction (31 December 2016), the mean age of people with type 2 diabetes was 68.7 ± 12.6 years. Less than half of these were female (43.9%). There were 3207 in the 25-44 years age category (12%), 12 110 aged 45-64 years (26%), 6407 aged 65-74 years (25%), 3639 aged 75-84 years (14%) and 817 aged 85 years or older (3.1%). Mean BMI was 31.72 (6.44) kg/m². Mean systolic blood pressure (BP) was 138.5 (17.4) mmHg and diastolic BP was 81.0 (10.7) mmHg. Ethnicity was identified in 85.5% of the diabetes cohort. People of Asian (6.6%) and Black (3.3%) ethnicity were representative of the general RCGP RSC population (5.8% and 3.5%, respectively).

The majority of the type 2 diabetes cohort (n = 26 180; 98.9%) had at least one measurement of BMI. The mean BMI was 30.7 ± 6.4 kg/m². The mean HbA1c measurement (most recent) for the cohort was 57.5 ± 16.5 mmol/mol [7.4% ± 1.5%]. The mean systolic BP was 139.0 (median: 138.0 [IQR: 129-148]) mmHg and the mean diastolic BP was 81.4 (median: 80.0 [IQR: 75-88]) mmHg. The mean estimated glomerular filtration rate was 81.4 (median: 82.2 [IQR: 68.2-94.9]) ml/min.

The numbers and percentages of individuals in the nine transition groups over the first 12 months were: A->A, n = 12 559 (48%); A->B, n = 1439 (5.5%); A->C, n = 388 (1.5%); B->A, n = 3822 (14.6%); B->B, n = 898 (3.4%); B->C, n = 451 (1.7%); C->A, n = 4223 (16.1%); C->B, n = 1736 (6.6%); and C->C, n = 664 (2.5%). The numbers and percentages of individuals in the five glycaemic variability groups were: score 0-20, n = 13 477 (51.5%); score 21-40, n = 4929 (18.8%); score 41-60, n = 4208 (16.1%); score 61-80, n = 2557 (9.8%); and score 81-100, n = 1009 (3.9%).

3.2 Major adverse cardiovascular events

From the 26 180 patients in the cohort, 4179 MACE were recorded prior to the diagnosis of type 2 diabetes. Within the first 12 months after diagnosis of diabetes, there were a further 1457 MACE, and from 12 months after the diagnosis of diabetes onwards, there were 2300 MACE. The median time to MACE after diagnosis was 635 (IQR 164-1539) days. Inclusive within the category of MACE was MI, of which there were 1424 events prior to the diagnosis of diabetes, 289 within 12 months of the diagnosis of diabetes, and a further 513 after 12 months of diabetes. The cumulative burden of disease in the cohort, at the end of follow-up, is shown in Table 1.

The modelling of the effect of change from baseline HbA1c to 12-month HbA1c suggested that those who moved from the highest category to the lowest category (C->A) had a reduced hazard for subsequent MACE (Table 2).

Increased glycaemic variability was associated with an increased risk of MACE; this risk increased across the categories of glycaemic variability, but was significant only for those in the highest category (P < .001). The modelling of the effect of change from baseline HbA1c to 12-month HbA1c suggested that those who moved from the highest category to the lowest category (C->A) had a reduced hazard for subsequent MACE (Table 2).

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DISCUSSION

We have shown that improvement of glycaemic control, within the first year following diagnosis of type 2 diabetes, was associated with a reduced risk of subsequent MACE. Furthermore, we found that the greatest glycaemic variability (i.e. after the initial 12 months of glycaemic trajectory) also conferred a higher risk of MACE. These data are in keeping with the concept of a legacy effect, but with subsequent glycaemic variability contributing to an individual’s risk profile.

4.1 | Trajectory/metabolic memory

Following the landmark trials of glucose regulation in type 2 diabetes (UKPDS, The Action to Control Cardiovascular Risk in Diabetes Study [ACCORD], Veterans Affairs Diabetes Trial [VADT], Action in Diabetes and Vascular Disease: Preterax and Diamicron MR Controlled Evaluation [ADVANCE]), trajectories of control have gained prominence. The effect of the metabolic memory may depend on the absolute reduction in HbA1c, the achievement below a ‘threshold value’, the speed at which that is obtained, and/or the duration for which it is maintained. Our data, of benefits associated with early control (by 1 year after diagnosis), concord with those of other groups.

Conversely, glycaemic control attained later than 1 year may lose benefit. Correction of high HbA1c over a prolonged period of 4-5 years was associated with either no benefit or worse outcomes. Delaying the institution of glycaemic control until several years after diagnosis of type 2 diabetes is also of limited benefit to macrovascular reduction. A randomized controlled trial (RCT) of multifactorial cardiovascular risk management (including normalization of HbA1c from 8.0% to 6.8%), found little benefit in type 2 diabetes of (mean) 8.5 years’ duration, although this study was underpowered, as the HbA1c in the control group also decreased (to 7.1%). Absence of benefit was also reported in ADVANCE, VADT and ACCORD, with
durations of diabetes of 7.9-11.5 years. Analysis of 10- versus 15-year follow-up of VADT suggests that the metabolic memory in intensively treated patients is time-limited, with any effect dissipating over prolonged follow-up. The duration of follow-up in our data was 12 years, and it might be that benefits of early control would become blunted thereafter.

4.2 Variability

HbA1c variability has been increasingly recognized as an adverse prognosticator in type 2 diabetes. Our data show that this effect was independent of the baseline HbA1c and trajectory of HbA1c change over the first year. Long-term variability, based on HbA1c, may represent poor adherence to dietary and pharmacological regimes, rather than having the same interpretation as short-term (daily) variability. Even so, ADVANCE showed that in a RCT (with the ability to monitor treatment adherence), greater HbA1c variability remained associated with an increased risk of vascular events and mortality.

HbA1c variability may be a surrogate for the overall quality of care. Therefore, we adjusted for baseline cardiovascular risk factors, although we did not have data for the transition of these factors over the follow-up period. Alternatively, the adverse effects of glycaemic variability may relate to sympathoadrenal activation from hypoglycaemia, causing myocardial irritability. However, mortality risk persists for months after hypoglycaemia. Hypoglycaemia, although considered under-reported, is highly prevalent among people with type 2 diabetes, particularly in late disease. Visit-to-visit variability in fasting glucose has also been linked to all-cause mortality.

We found that younger adults, aged 25-44 years, were more probable to remain in the HbA1c category over the first year. This is concerning, as these patients may be more probable to benefit from a later pay-off from a legacy effect. This observation should be explored in future studies.

There is currently no standardized definition for HbA1c variability. Studies have expressed variability as the SD or coefficient of variation for all HbA1c measurements. However, in view of the progressive nature of type 2 diabetes and the tendency of HbA1c to increase over time, the SD would be inflated relative to the mean without representing fluctuation in HbA1c. We considered an absolute change in HbA1c of at least 0.5% as being clinically meaningful, but modified the method of Forbes et al. to give greater statistical weight to more recent variability in HbA1c.

4.3 Strengths and limitations

The large size of the cohort and the robust nature of the diabetes case finding and classification algorithm are major advantages of our approach. Most type 2 diabetes in the UK is treated within primary care, meaning that it is improbable that any measurements of HbA1c would not have been recorded, and thus the results are translatable to the wider population. Other studies of variability have been made from secondary care, with the potential for bias. Only one other study has used primary care data in newly diagnosed type 2 diabetes, and no studies have addressed the twin components of early control plus later variability, using primary care data.

Relationships of glycaemic patterns with mortality were not possible because of incomplete mortality data. Our primary endpoint was MACE, identified from primary care records. It is possible that some events, resulting in acute hospitalization, may not have been entered into the primary care record. However, we have shown that CVD is well recorded in the RCGP RSC. Analysis of long-term VADT data showed that intensive glycaemic control reduces the relative risk of non-fatal MI and coronary heart disease events, but without an effect on mortality. However, those analyses were not restricted to early onset of intensive therapy. We did not include diabetes medications as a covariate, as the relationship between HbA1c and diabetes medications is not unidirectional. The BMI may modify the relationship between HbA1c and mortality in patients with type 2 diabetes and so we adjusted for BMI at baseline, but BMI was not recorded sufficiently frequently to allow time-varying adjustment. Prediagnosis glucose exposure (duration and concentration) may have a bearing on subsequent outcomes, but—as with most studies of type 2 diabetes—we were unable to quantify the exposure to prolonged hyperglycaemia prior to the diagnosis of diabetes, although we did adjust for initial HbA1c.

Newer glucose-lowering agents (sodium-glucose co-transporter-2 inhibitors [SGLT2is] and glucagon-like peptide-1 receptor agonists [GLP-1 RAs]) have the added attribute of cardiovascular risk reduction and may also reduce glycaemic variability. The use of SGLT2is and GLP-1RAs, at the time of data extraction (31 December 2016), was less than 5% among individuals with type 2 diabetes in our cohort. The numbers in some of the first-year transition groups were less than 2% of the total cohort, consequently confidence intervals were wide. We excluded patients who did not have HbA1c measurements at both diagnosis and 1 year after (±3 months either side of diagnosis and at 1-year follow-up), plus at least five measurements of HbA1c thereafter. This may affect interpretation as to whether these results are representative of most patients with type 2 diabetes.

For the variability analysis, we used time-updated analyses, as greater weight is placed on more recent glycaemic exposure; however, with this approach there is a greater risk of reverse causality. Nevertheless, analysis of data using a time-fixed approach did not appreciably change the outcome. Transitioning from C to A categories within the first year was associated with reduced hazard for MACE, and therefore further research into the value of early intensification is needed, to establish whether benefits are therapy-related, and the appropriateness for more frail individuals.

In conclusion, we found that (i) transitioning to an HbA1c of less than 7.5% (=59 mmol/mol) in the first year after diagnosis of type 2 diabetes was associated with reduced MACE, and (ii) thereafter, lack of substantial variability also lowered the risk for MACE. Our findings support the concept that effort must be made to achieve rapid metabolic normalization after the diagnosis of diabetes in those with a low propensity to hypoglycaemia.
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CONFLICT OF INTEREST
MBW has received investigator-led grant funding from AstraZeneca, Sanofi, Eli Lilly and speaker fees from AstraZeneca and MSD. AMcG has received research funding from Eli Lilly, AstraZeneca, Boehringer-Ingelheim and Pfizer. NM has received fees for serving as a speaker, a consultant or an advisory board member for Allergan, Bristol-Myers Squibb-Astra Zeneca, GlaxoSmithKline, Eli Lilly, Lifescan, MSD, Metronic, Novartis, Novo Nordisk, Pfizer, Sankio, Sanofi, Roche, Servier and Takeda. SdeL has held grants through his University for investigator-led studies from AstraZeneca, Eli Lilly Company (which funded academic time including part of WHs to work on this project), GlaxoSmithKline, MSD, Takeda and Novo Nordisk Limited. JM is an employee of Eli Lilly Company. WH, JvV, FF, MJ, UH report no conflicts of interest.

AUTHOR CONTRIBUTIONS
MBW: conceptualization, formal analysis, writing original draft. MJ: methodology, formal analysis. WH: formal analysis, writing – review & editing. AMcG: formal analysis, writing – review & editing. UH: formal analysis, writing – review & editing. JvV: methodology, formal analysis. FF: data curation, project administration. JM: project administration; writing original draft; writing – review & editing. NM: conceptualization, data interpretation, writing original draft. SdeL: conceptualization, writing original draft, oversight and leadership responsibility for the research activity planning and execution.

PEER REVIEW
The peer review history for this article is available at https://publons.com/publon/10.1111/dom.14705.

DATA AVAILABILITY STATEMENT
The data that support the findings of this study are available on request from the corresponding author. The data are not publicly available due to privacy or ethical restrictions.

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**SUPPORTING INFORMATION**

Additional supporting information may be found in the online version of the article at the publisher’s website.

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