Clinical Targets for Continuous Glucose Monitoring Data Interpretation: Recommendations From the International Consensus on Time in Range

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Improvements in sensor accuracy, greater convenience and ease of use, and expanding reimbursement have led to growing adoption of continuous glucose monitoring (CGM). However, successful utilization of CGM technology in routine clinical practice remains relatively low. This may be due in part to the lack of clear and agreed-upon glycemic targets that both diabetes teams and people with diabetes can work toward. Although unified recommendations for use of key CGM metrics have been established in three separate peer-reviewed articles, formal adoption by diabetes professional organizations and guidance in the practical application of these metrics in clinical practice have been lacking. In February 2019, the Advanced Technologies & Treatments for Diabetes (ATTD) Congress convened an international panel of physicians, researchers, and individuals with diabetes who are expert in CGM technologies to address this issue. This article summarizes the ATTD consensus recommendations for relevant aspects of CGM data utilization and reporting among the various diabetes populations.

Adoption of continuous glucose monitoring (CGM), which includes both real-time CGM (rtCGM) and intermittently scanned CGM (isCGM), has grown rapidly over the past few years as a result of improvements in sensor accuracy, greater convenience and ease of use, and expanding reimbursement. Numerous studies have demonstrated significant clinical benefits of CGM use in people with diabetes regardless of insulin delivery method (1–15). In many countries, the benefits and utility of CGM are now recognized by national and international medical organizations for individuals with insulin-requiring diabetes and/or those at risk for hypoglycemia (16–21). However, despite increased CGM adoption (22,23), successful utilization of CGM data in routine clinical practice remains relatively low. This may be due in part to the lack of clear and agreed-upon glycemic targets toward which both diabetes teams and people with diabetes can work.

In 2012 the Helmsley Charitable Trust sponsored the first expert panel to recommend the standardization of CGM metrics and CGM report visualization (24). This was followed by a series of CGM consensus statements refining the core CGM metrics, but the conclusions were never in alignment. In 2017, several articles supported use of systematic approaches to CGM data evaluation (18–20). To date, the key CGM metrics remain as unified recommendations in three separate peer-reviewed articles, yet formal adoption by diabetes professional organizations and

This international consensus report has been endorsed by the American Diabetes Association, American Association of Clinical Endocrinologists, American Association of Diabetes Educators, European Association for the Study of Diabetes, Foundation of European Nurses in Diabetes, International Society for Pediatric and Adolescent Diabetes, JDRF, and Pediatric Endocrine Society.

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guidance in the practical application of these metrics in clinical practice have been lacking (19).

In February 2019, the Advanced Technologies & Treatments for Diabetes (ATTD) Congress convened an international panel of individuals with diabetes and clinicians and researchers with expertise in CGM. Our objective was to develop clinical CGM targets to supplement the currently agreed-upon metrics for CGM-derived times in glucose ranges (within target range, below target range, above target range) in order to provide guidance for clinicians, researchers, and individuals with diabetes in using, interpreting, and reporting CGM data in routine clinical care and research. Importantly, in order to make the recommendations generalizable and comprehensive, the consensus panel included individuals living with diabetes and had international representation from physicians and researchers from all geographic regions.

The panel was divided into subgroups to review literature and provide recommendations for relevant aspects of CGM data utilization and reporting among the various diabetes populations. Long-term trials demonstrating how CGM metrics relate to and/or predict clinical outcomes have not been conducted, and many of the published reports assessed here are not at the highest evidence level (25). However, there is suggestive evidence from a number of recent studies, including a cross-sectional study correlating current retrospective 3-day time in target range with varying degrees of diabetes retinopathy (26) and an analysis of the 7-point self-monitored blood glucose (SMBG) data from the Diabetes Control and Complications Trial (DCCT) (27), showing correlations of time in target range (70–180 mg/dL [3.9–10.0 mmol/L]) with diabetes complications. Relationships between time in target range and A1C (26,27) and number of severe and nonsevere hypoglycemic events (28–32) have also been observed. Recommendations from each subgroup were presented to the full panel and voted upon. This article summarizes the consensus recommendations and represents the panel members’ evaluation of the issues.

NEED FOR METRICS BEYOND A1C

A1C is currently recognized as the key surrogate marker for the development of long-term diabetes complications in people with type 1 and type 2 diabetes and has been used as the primary end point for many CGM studies (1,3,4,6,33,34). While A1C reflects average glucose over the last 2–3 months, its limitation is the lack of information about acute glycemic excursions and the acute complications of hypo- and hyperglycemia. A1C also fails to identify the magnitude and frequency of intra- and interday glucose variation (35,36). Moreover, certain conditions such as anemia (37), hemoglobinopathies (38), iron deficiency (39), and pregnancy (40) can confound A1C measurements. Importantly, as reported by Beck et al. (41), the A1C test can fail at times to accurately reflect mean glucose even when none of those conditions are present. Despite these limitations, A1C is the only prospectively evaluated tool for assessing the risk for diabetes complications, and its importance in clinical decision making should not be undervalued. Rather, the utility of A1C is further enhanced when used as a complement to glycemic data measured by CGM.

Unlike A1C measurement, use of CGM allows for the direct observation of glycemic excursions and daily profiles, which can inform on immediate therapy decisions and/or lifestyle modifications. CGM also provides the ability to assess glucose variability and identify patterns of hypo- and hyperglycemia. However, potential drawbacks of CGM use include the need to be actively used in order to be effective; that it may induce anxiety;
that it may have accuracy limitations, particularly with the delay in registering blood glucose changes in dynamic situations; and that it can provoke allergies. Another limitation of CGM is that this technology is not yet widely available in several regions of the world.

Effective use of CGM data to optimize clinical outcomes requires the user to interpret the collected data and act upon them appropriately. This requires 1) common metrics for assessment of CGM glycemic status, 2) graphical visualization of the glucose data and CGM daily profile, and 3) clear clinical targets.

**STANDARDIZATION OF CGM METRICS**

In February 2017, the ATTD Congress convened an international panel of expert clinicians and researchers to define core metrics for assessing CGM data (18) (Table 1).

The list of core metrics has now been streamlined for use in clinical practice based on the expert opinion of this international consensus group (18). Of the 14 core metrics, the panel selected that 10 metrics that may be most useful in clinical practice (Table 2).

Fundamental to accurate and meaningful interpretation of CGM is ensuring that adequate glucose data are available for evaluation. As shown in studies, >70% use of CGM over the most recent 14 days correlates strongly with 3 months of mean glucose, time in ranges, and hyperglycemia metrics (42,43). In individuals with type 1 diabetes, correlations are weaker for hypoglycemia and glycemic variability; however, these correlations have not been shown to increase with longer sampling periods (43). Longer CGM data collection periods may be required for individuals with more variable glycemic control (e.g., 4 weeks of data to investigate hypoglycemia exposure).

**TIME IN RANGES**

The development of blood glucose testing provided individuals with diabetes the ability to obtain immediate information about their current glucose levels and adjust their therapy accordingly. Over the past decades, national and international medical organizations have been successful in developing, harmonizing, and disseminating standardized glycemic targets based on risk for acute and chronic complications. CGM technology greatly expands the ability to assess glycemic control throughout the day, presenting critical data to inform daily treatment decisions and quantifying time below, within, and above the established glycemic targets.

Although each of the core metrics established in the 2017 ATTD consensus conference (18) provides important information about various aspects of glycemic status, it is often impractical to assess and fully utilize many of these metrics in real-world clinical practices. To streamline data interpretation, the consensus panel identified “time in ranges” as a metric of glycemic control that provides more actionable information than A1C alone. The panel agreed that establishing target percentages of time in the various glycemic ranges with the ability to adjust the percentage cut points to address the specific needs of special diabetes populations (e.g., pregnancy, high-risk) would facilitate safe and effective therapeutic decision making within the parameters of the established glycemic goals.

The metric includes three key CGM measurements: percentage of readings and time per day within target glucose range (TIR), time below target glucose range (TBR), and time above target glucose range (TAR) (Table 3). The primary goal for effective and safe glucose control is to increase the TIR while reducing the TBR. The consensus group agreed that expressing time in the various ranges can be done as the percentage (%) of CGM readings, average hours and minutes spent in each range per day, or both, depending on the circumstances.

It was agreed that CGM-based glycemic targets must be personalized to meet the needs of each individual with diabetes. In addition, the group reached consensus on glycemic cutpoints (a target range of 70–180 mg/dL [3.9–10.0 mmol/L] for individuals with type 1 diabetes and type 2 diabetes and 63–140 mg/dL [3.5–7.8 mmol/L] during pregnancy, along with a set of targets for the time per day [% of CGM readings or minutes/

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**Table 1—Standardized CGM metrics**

2017 international consensus on CGM metrics (18)

<table>
<thead>
<tr>
<th>1. Number of days CGM worn</th>
<th>14 days</th>
</tr>
</thead>
<tbody>
<tr>
<td>2. Percentage of time CGM is active</td>
<td>70%</td>
</tr>
<tr>
<td>3. Mean glucose</td>
<td>7.8 mmol/L</td>
</tr>
<tr>
<td>4. Estimated A1C</td>
<td>Type 1 diabetes: 7.5% Type 2 diabetes: 7.0%</td>
</tr>
<tr>
<td>5. Glycemic variability (%CV or SD)</td>
<td>1.2</td>
</tr>
<tr>
<td>6. Time &gt;250 mg/dL (13.9 mmol/L)</td>
<td>15%</td>
</tr>
<tr>
<td>7. Time &gt;180 mg/dL (10.0 mmol/L)</td>
<td>10%</td>
</tr>
<tr>
<td>8. Time 70–180 mg/dL (3.9–10.0 mmol/L)</td>
<td>50%</td>
</tr>
<tr>
<td>9. Time &lt;70 mg/dL (&lt;3.9 mmol/L)</td>
<td>15%</td>
</tr>
<tr>
<td>10. Time &lt;54 mg/dL (&lt;3.0 mmol/L)</td>
<td>5%</td>
</tr>
<tr>
<td>11. LBGI and HBGI (risk indices)</td>
<td>19%</td>
</tr>
<tr>
<td>12. Episodes (hypoglycemia and hyperglycemia)</td>
<td>15 min</td>
</tr>
<tr>
<td>13. Area under the curve</td>
<td>1.2</td>
</tr>
<tr>
<td>14. Time blocks (24-h, day, night)</td>
<td>80%</td>
</tr>
</tbody>
</table>

**Use of Ambulatory Glucose Profile (AGP) for CGM report**

CV, coefficient of variation; LBGI, low blood glucose index; HBGI, high blood glucose index.

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**Table 2—Standardized CGM metrics for clinical care: 2019**

<table>
<thead>
<tr>
<th>1. Number of days CGM worn (recommend 14 days)</th>
<th>14 days</th>
</tr>
</thead>
<tbody>
<tr>
<td>2. Percentage of time CGM is active (recommend 70% of data from 14 days)</td>
<td>70%</td>
</tr>
<tr>
<td>3. Mean glucose</td>
<td>7.8 mmol/L</td>
</tr>
<tr>
<td>4. Glucose management indicator (GMI)</td>
<td>15%</td>
</tr>
<tr>
<td>5. Glycemic variability (%CV) target ≤36% (90)*</td>
<td>19%</td>
</tr>
<tr>
<td>6. Time above range (TAR): % of readings and time &gt;250 mg/dL (&gt;13.9 mmol/L)</td>
<td>Level 2</td>
</tr>
<tr>
<td>7. Time above range (TAR): % of readings and time 181–250 mg/dL (10.1–13.9 mmol/L)</td>
<td>Level 1</td>
</tr>
<tr>
<td>8. Time in range (TIR): % of readings and time 70–180 mg/dL (3.9–10.0 mmol/L)</td>
<td>In range</td>
</tr>
<tr>
<td>9. Time below range (TBR): % of readings and time 54–69 mg/dL (3.0–3.8 mmol/L)</td>
<td>Level 1</td>
</tr>
<tr>
<td>10. Time below range (TBR): % of readings and time &lt;54 mg/dL (&lt;3.0 mmol/L)</td>
<td>Level 2</td>
</tr>
</tbody>
</table>

**Use of Ambulatory Glucose Profile (AGP) for CGM report**

CV, coefficient of variation. *Some studies suggest that lower %CV targets (<33%) provide additional protection against hypoglycemia for those receiving insulin or sulfonylureas (45,90,91).
hours) individuals with type 1 diabetes and type 2 diabetes (Table 3) and women during pregnancy (Table 4) should strive to achieve. It should be noted that premeal and postprandial SMBG targets remain for diabetes in pregnancy (44), in addition to the new CGM TIR targets for overall glycemia.

Although the metric includes TIR, TBR, and TAR, achieving the goals for both TBR and TIR would result in reduced time spent above range and thereby improve glycemic control. However, some clinicians may choose to target the reduction of the high glucose values and minimize hypoglycemia, thereby arriving at more time in the target range. In both approaches, the first priority is to reduce TBR to target levels and then address TIR or TAR targets.

Note that for people with type 1 diabetes, the targets are informed by the ability to reach the targets with hybrid closed-loop therapy (11), the first example of which is now commercially available with several more systems in final stages of testing. Importantly, recent studies have shown the potential of reaching these targets with CGM in individuals using multiple daily injections (6). In type 2 diabetes, there is generally less glycemic variability and hypoglycemia than in type 1 diabetes (45). Thus, people with type 2 diabetes can often achieve more time in the target range while minimizing hypoglycemia (4). As demonstrated by Beck et al. (4), individuals with type 2 diabetes increased their TIR by 10.3% (from 55.6% to 61.3%) after 24 weeks of CGM use with slight reductions in TBR. Most recently, the beneficial effects of new medications, such as sodium–glucose cotransporter 2 agents have helped individuals with type 1 diabetes increase TIR (46–48). Targets for type 1 diabetes and type 2 diabetes were close enough to combine into one set of targets, outside of pregnancy.

Another way to visualize the CGM-derived targets for the four categories of diabetes is shown in Fig. 1, which displays and compares the targets for TIR (green), TBR (two categories in light and dark red), and TAR (two categories in yellow and orange). It becomes clear at a glance that there are different expectations for the various time in ranges relating to safety concerns and efficacy based on currently available therapies and medical practice.

### CLINICAL VALIDITY OF MEASURES

To fundamentally change clinical care with use of the new metrics, it would be important to demonstrate that the metrics relate to and predict clinical outcomes. In this regard, longer-term studies relating to time spent within specific CGM glycemic ranges, diabetes complications, and other outcomes are required. However, there is evidence from a number of recent studies that have shown correlations of TIR (70–180 mg/dL [3.9–10.0 mmol/L]) with diabetes complications (49,50) as well as a relationship between TIR and A1C (26,27). Although evidence regarding TIR for older and/or high-risk individuals is lacking, numerous studies have shown the elevated risk for hypoglycemia in these populations (51–56). Therefore, we have lowered the TIR target from
>70% to >50% and reduced TBR to <1% at <70 mg/dL (<3.9 mmol/L) to place greater emphasis on reducing hypoglycemia with less emphasis on maintaining target glucose levels (Table 3).

**Type 1 Diabetes and Type 2 Diabetes**

**Association With Complications**

Associations between TIR and progression of both diabetic retinopathy (DR) and development of microalbuminuria were reported by Beck et al. (50), using 7-point blood glucose profiles from the DCCT data set to validate the use of TIR as an outcome measure for clinical trials. Their analysis showed that the hazard rate for retinopathy progression increased by 64% for each 10% reduction in TIR. The hazard rate for microalbuminuria development increased by 40% for each 10% reduction in TIR. A post-hoc analysis of the same DCCT data set showed a link between glucose of 70–180 mg/dL (3.9–10.0 mmol/L) and an increased risk for severe hypoglycemia (57).

**Relationship Between TIR and A1C**

Analyses were conducted utilizing data-sets from four randomized trials encompassing 545 adults with type 1 diabetes who had central laboratory measurements of A1C (26). TIR (70–180 mg/dL [3.9–10.0 mmol/L]) of 70% and 50% strongly corresponded with an A1C of approximately 7% (53 mmol/mol) and 8% (64 mmol/mol), respectively. An increase in TIR of 10% (2.4 h per day) corresponded to a decrease in A1C of approximately 0.5% (5.0 mmol/mol); similar associations were seen in an analysis of 18 randomized controlled trials (RCTs) by Vigersky and McMahon (27) that included over 2,500 individuals with type 1 diabetes and type 2 diabetes over a wide range of ages and A1C levels (Table 5).

**Pregnancy**

During pregnancy, the goal is to safely increase TIR as quickly as possible, while reducing TAR and glycemic variability. Data from the first study of longitudinal CGM use in pregnancy demonstrated a 13-percentage point increase in TIR (43% to 56% TIR 70–140 mg/dL [3.9–7.8 mmol/L]) (58). TBR <50 mg/dL was reduced from 6% to 4%, although the higher TBR <70 mg/dL was high (13–15%) using older-generation sensors. With improved sensor accuracy, recent type 1 diabetes pregnancy studies report a lower threshold of <63 mg/dL (<3.5 mmol/L) for TBR and ≥63 mg/dL (≥3.5 mmol/L) for TIR (59,60). Data from Sweden, and the Continuous Glucose Monitoring in Women With Type 1 Diabetes in Pregnancy Trial (CONCEPTT) control group, report 50% TIR in the first trimester, improving to 60% TIR in the third trimester, reflecting contemporary antenatal care. Of note, these data confirm that the TBR <63 mg/dL (<3.5 mmol/L) recommendation of <4% is safely achievable, especially after the first trimester. Furthermore, 33% of women achieved the recommendation of 70% TIR 63–140 mg/dL (3.5–7.8 mmol/L) in the final (>34) weeks of pregnancy. Preliminary data suggest that closed-loop systems may allow pregnant women to safely achieve 70% TIR at an earlier (>24 weeks) stage of gestation (61,62). Law et al. (63) analyzed data from two early CGM trials (64,65) describing the associations between CGM measures and risk of large-for-gestational-age (LGA) infants. Taken together, the Swedish and CONCEPTT data confirm that a 5–7% higher
TIR during the second and third trimesters is associated with decreased risk of LGA and neonatal outcomes including macrosomia, shoulder dystocia, neonatal hypoglycemia, and neonatal intensive care admissions. More data are needed to define the clinical CGM targets for pregnant women with type 2 diabetes, who spend one-third less time hyperglycemic than women with type 1 diabetes and achieve TIR of 90% (58). Because of the lack of evidence on CGM targets for women with gestational diabetes mellitus (GDM) or type 2 diabetes in pregnancy, percentages of time spent in range, below range, and above range have not been included in this report. Recent data suggest that even more stringent targets (66) and greater attention to overnight glucose profiles may be required to normalize outcomes in pregnant women with GDM (63).

**Older and/or High-Risk Individuals With Diabetes**

Older and/or high-risk individuals with diabetes are at notably higher risk for severe hypoglycemia due to age, duration of diabetes, duration of insulin therapy, and greater prevalence of hypoglycemia unawareness (51–55). The increased risk of severe hypoglycemia is compounded by cognitive and physical impairments and other comorbidities (53,56). High-risk individuals include those with a higher risk of complications, comorbidity conditions (e.g., cognitive deficits, renal disease, joint disease, osteoporosis, fracture, and/or cardiovascular disease), and those requiring assisted care, which can complicate treatment regimens (56). Therefore, when setting glycemic targets for high-risk and/or elderly people, it is important to individualize and be conservative, with a strong focus on reducing the percentage of time spent <70 mg/dL (<3.9 mmol/L) and preventing excessive hyperglycemia.

**STANDARDIZATION OF CGM DATA PRESENTATION**

As noted above, in 2013 a panel of clinicians with expertise in CGM published recommendations for use of the Ambulatory Glucose Profile (AGP) as a template for data presentation and visualization. Originally created by Mazze et al. (67), the standardized AGP report was further developed by the International Diabetes Center and now incorporates all the core CGM metrics and targets along with a 14-day composite glucose profile as an integral component of clinical decision making (24). This recommendation was later endorsed at the aforementioned international consensus conference on CGM metrics (18) and is referenced as an example in the American Diabetes Association 2019 “Standards of Medical Care in Diabetes” (16) and in an update to the American Association of Clinical Endocrinologists consensus on use of CGM (68). The AGP report, in slightly modified formats, has been adopted by most of the CGM device manufacturers in their downloadable software. An example of the AGP report, updated to incorporate targets, is presented in Fig. 2. In the AGP report, glucose ranges are defined as "Very High" (Level 2), “High” (Level 1), “Low” (Level 1), and "Very Low" (Level 2). An “mmol/L” version is provided in Supplementary Fig. 1.

There is a general consensus that a useful CGM report is one that can be understood by clinicians and people with diabetes. While there may be some terms (e.g., glucose variability) that are less familiar to many people with diabetes, a single-page report that the medical team can review and file in the electronic medical record and that can be used as a shared decision-making tool with people with diabetes was considered to be of value (69–72). More detailed reports (e.g., adjustable data ranges, detailed daily reports) should remain available for individualized review by or with people with diabetes.

**Clinical Application of Time in Ranges**

Despite its demonstrated value, clinical utilization of CGM data has remained suboptimal. Although time constraints and reimbursement issues are clearly obstacles, clinician inexperience in data interpretation and lack of standardization software for visualization of CGM data have also played a role (73). The proposed standardized report enables clinicians to readily identify important metrics such as the percentage of time spent within, below, and above each individual’s target range, allowing for greater personalization of therapy through shared decision making.

Using the standardized report, the clinician can also address glucose variability (e.g., the coefficient of variation

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**Table 5—Estimate of A1C for a given TIR level based on type 1 diabetes and type 2 diabetes studies**

<table>
<thead>
<tr>
<th>TIR (70–180 mg/dL)</th>
<th>A1C, % (mmol/mol)</th>
<th>95% CI for predicted A1C values, %</th>
</tr>
</thead>
<tbody>
<tr>
<td>20% (3.9–10.0 mmol/L)</td>
<td>9.4 (79)</td>
<td>(8.0, 10.7)</td>
</tr>
<tr>
<td>30%</td>
<td>8.9 (74)</td>
<td>(7.6, 10.2)</td>
</tr>
<tr>
<td>40%</td>
<td>8.4 (68)</td>
<td>(7.1, 9.7)</td>
</tr>
<tr>
<td>50%</td>
<td>7.9 (63)</td>
<td>(6.6, 9.2)</td>
</tr>
<tr>
<td>60%</td>
<td>7.4 (57)</td>
<td>(6.1, 8.8)</td>
</tr>
<tr>
<td>70%</td>
<td>7.0 (53)</td>
<td>(5.6, 8.3)</td>
</tr>
<tr>
<td>80%</td>
<td>6.5 (48)</td>
<td>(5.2, 7.8)</td>
</tr>
<tr>
<td>90%</td>
<td>6.0 (42)</td>
<td>(4.7, 7.3)</td>
</tr>
</tbody>
</table>

Every 10% increase in TIR = −0.5% (5.5 mmol/mol) A1C reduction

<table>
<thead>
<tr>
<th>TIR (70–180 mg/dL)</th>
<th>A1C, % (mmol/mol)</th>
</tr>
</thead>
<tbody>
<tr>
<td>20% (3.9–10.0 mmol/L)</td>
<td>10.6 (92)</td>
</tr>
<tr>
<td>30%</td>
<td>9.8 (84)</td>
</tr>
<tr>
<td>40%</td>
<td>9.0 (75)</td>
</tr>
<tr>
<td>50%</td>
<td>8.3 (67)</td>
</tr>
<tr>
<td>60%</td>
<td>7.5 (59)</td>
</tr>
<tr>
<td>70%</td>
<td>6.7 (50)</td>
</tr>
<tr>
<td>80%</td>
<td>5.9 (42)</td>
</tr>
<tr>
<td>90%</td>
<td>5.1 (32)</td>
</tr>
</tbody>
</table>

Every 10% increase in TIR = −0.8% (8.7 mmol/mol) A1C reduction

The difference between findings from the two studies likely stems from differences in number of studies analyzed and subjects included (RCTs with subjects with type 1 diabetes vs. RCTs with subjects with type 1 or type 2 diabetes with CGM and SMBG).
**AGP Report**

**GLUCOSE STATISTICS AND TARGETS**

<table>
<thead>
<tr>
<th>Date Range</th>
<th>Days Active</th>
<th>% Time CGM Active</th>
</tr>
</thead>
<tbody>
<tr>
<td>26 Feb 2019–10 Mar 2019</td>
<td>13</td>
<td>99.9%</td>
</tr>
</tbody>
</table>

**Glucose Ranges**

- **Target Range** 70–180 mg/dL
- **Greater than 70%** (16h 48min)
- **Less than 4%** (58min)
- **1%** (14min)
- **Less than 25%** (6h)
- **Less than 5%** (1h 12min)

Each 5% increase in time in range (70–180 mg/dL) is clinically beneficial.

**Average Glucose** 173 mg/dL

**Glucose Management Indicator (GMI)** 7.6%

**Glucose Variability** 49.5%

Defined as percent coefficient of variation (%CV); target ≤36%

**TIME IN RANGES**

- **Very High** (>250 mg/dL) 20% (4h 48min)
- **High** (181–250 mg/dL) 23% (5h 31min)
- **Target Range** (70–180 mg/dL) 47% (11h 17min)
- **Low** (54–69 mg/dL) 4% (58min)
- **Very Low** (<54 mg/dL) 6% (1h 26min)

**AMBULATORY GLUCOSE PROFILE (AGP)**

AGP is a summary of glucose values from the report period, with median (50%) and other percentiles shown as if occurring in a single day.

**DAILY GLUCOSE PROFILES**

Each daily profile represents a midnight-to-midnight period.

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**Figure 2**—Ambulatory Glucose Profile.
Goal Setting

Numerous studies have demonstrated the clinical benefits of early achievement of near-normal glycemic control in individuals with type 1 diabetes and type 2 diabetes (77–83). However, when advising people with diabetes, goal-setting must be collaborative and take into account the individual needs/capabilities of each patient and start with the goals that are most achievable. An early study by DeWalt et al. (84) found that setting small, achievable goals not only enhances people’s ability to cope with their diabetes, but that people with diabetes who set and achieved their goals often initiated additional behavioral changes on their own. One approach to consider is the SMART goal (Specific, Measurable, Achievable, Relevant, Time-bound) intervention, which is directly applicable to setting targets for time in ranges. First described by Lawlor and Hornyak in 2012 (85), this approach incorporates four key components of behavioral change relevant to goal setting.

1. The goal is specific and defines exactly what is to be achieved.
2. The goal is measurable and there is tangible evidence when it has been achieved.
3. The goal is achievable but stretches the patient slightly so that he/she feels challenged.
4. The goal should be attainable over a short period of time.

Effective goals should utilize CGM data to identify specific instances for the patient to take measurable action to prevent hypoglycemia. Although analysis of the AGP reports provides an opportunity for meaningful discussion, individuals should be counseled to look at patterns throughout the day to see when low glucose events are occurring and make adjustments in their therapy to reduce these events.

When applying the CGM metrics in clinical practice, it may be more meaningful and motivating to communicate to people with diabetes the importance of working to reduce the time spent below the threshold for hypoglycemia (%CV metric) (74) or use the glucose management indicator (GMI) metric (75) to discuss the possible discrepancies noted in glucose exposure derived from CGM data versus the individual’s laboratory-measured A1C (41,76). With appropriate educational materials, time, and experience, clinicians will develop a systematic approach to CGM data analysis and the most effective ways to discuss the data with patients in person or remotely.

CONCLUSIONS

Use of CGM continues to expand in clinical practice. As a component of diabetes self-management, daily use of CGM provides the ability to obtain immediate feedback on current glucose levels as well as direction and rate of change in glucose levels. This information allows people with diabetes to optimize dietary intake and exercise, make informed therapy decisions regarding mealtime and correction of insulin dosing, and, importantly, react immediately and appropriately to mitigate or prevent acute glycemic events (87–89). Retrospective analysis of CGM data, using standardized data management tools such as the AGP, enables clinicians and people with diabetes to work collaboratively in identifying problem areas and then set achievable goals (70–72). We conclude that, in clinical practice, time in ranges (within target range, below range, above range) are both appropriate and useful as clinical targets and outcome measurements that complement A1C for a wide range of people with diabetes and that the target values specified in this article should be considered an integral component of CGM data analysis and day-to-day treatment decision making.

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