

Supplementary Data

Supplementary S1 Table. Continuous Glucose Analysis – Definitions and measures of glucose series

<i>Parameter</i>	<i>Definition</i>	<i>Equation</i>	<i>Reference</i>
Statistical measures and temporal events			
MEAN_I	Mean glucose over time interval I I: 24 h Before Breakfast/ Lunch/ Dinner After Breakfast/ Lunch/ Dinner Day Night Early Morning	$MEAN_I = \frac{\sum_{i=1}^{N_I} g_i}{N_I}$	[1]
AUC180/ iAUC	<i>AUC180</i> - Percentage Area under glucose curve above target range (180 mg/dl) <i>iAUC</i> - includes all area below the glucose curve and above the fasting concentration.	$\frac{\sum_{i=1}^{N_I-1} (g_i + g_{i+1} - 360) \cdot \Delta t / 2, g_i \geq 180}{\sum_{i=1}^{N_I-1} (g_i + g_{i+1} - 2 \cdot g_{fasting}) \cdot \Delta t / 2, g_i \geq g_{fasting}}$	[2]
Max_I	Max glucose level within 4 h after meal ingestion I: Breakfast/ Lunch/ Dinner	$\max_{4h} [g_0 \dots g_{N_I}]$	

Glucose variability and risk assessment			
SD	Standard Deviation	$SD = \sqrt{\frac{\sum_{i=1}^{N_I} (g_i - Mean_I)^2}{N_I - 1}}$	[1]
% CV	Coefficient of Variation	$CV = \frac{SD_I \cdot 100}{Mean_I}$	[1]
M-value	weighted SD	$M = \frac{\sum_{i=1}^{N_I} \left 10 \cdot \log \frac{g_i}{120} \right ^3}{N_I}$	[3]
MAGE	Mean Amplitude of Glycemic Excursion	$MAGE = MEAN \lambda $ <p>where λ = change in glucose levels over each excursion from peak to nadir (or nadir to peak)</p> <p>and</p> $\lambda > SD_{24}$	[4]
MODD	Mean of Daily Differences	$MODD = \frac{\sum_{i=1}^{N_I} (g_i - g_{i-f \cdot 24}) }{N_I}$	[5]
CONGA_n	SD of the difference between values obtained exactly n hours apart n = 1, 2, 4, etc.	$CONGA_n = \sqrt{\frac{\sum_{i=1}^{N_I} (D_t - \bar{D})^2}{N_I - 1}}$	[6]

		<p>where</p> $D_t = g_i - g_{i-f \cdot n}$ <p>and</p> $\bar{D} = \frac{\sum_{i=1}^{N_I} D_t}{N_I}$	
BGRI	<p>LBGI - Low Blood Glucose Index</p> <p>HBGI - High Blood Glucose Index</p> <p>BGRI - Blood glucose risk index</p>	$f(g_i) = 1.509 \cdot [(\ln(g_i))^{1.084} - 5.381]$ $r(g_i) = 10 \cdot f(g_i)^2$ $rl(g_i) = r(g_i) \text{ if } f(g_i) < 0 \text{ and } 0 \text{ o.w}$ $rh(g_i) = r(g_i) \text{ if } f(g_i) > 0 \text{ and } 0 \text{ o.w}$ $LBGI = \frac{1}{N_I} \sum_{i=1}^{N_I} rl(g_i)$ $HBGI = \frac{1}{N_I} \sum_{i=1}^{N_I} rh(g_i)$ $BGRI = LBGI + HBGI$	[7]

g_i = Continuous glucose monitoring (CGM) glucose time series – glucose reading i

N_I = number of glucose readings for a given time interval I

Δt = time interval between adjacent samples (min)

f = sampling frequency (samples per hour)

References

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