THE LANCET Digital Health

Supplementary appendix

This appendix formed part of the original submission and has been peer reviewed. We post it as supplied by the authors.

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Supplementary Materials

1 CONTROL ALGORITHM DESIGN

1.1 EXMPC CONTROLLER DESIGN

The schematic of the exMPC used in this study can be found in Figure 1 in [1].

The exercise-enabled model-predictive control (exMPC) described here is designed to be used in hybrid mode, which requires the user to announce meals to the system. The exMPC consists of a process model (i.e. mathematical model), a reference trajectory, and an optimization tool to calculate the next insulin delivery rate. The process model is a mathematical description of glucose and insulin metabolism. It is defined by five differential equations in this study and is further described in multiple publications by Resalat and colleagues [1-4]. The ALPHA algorithm [3] adapts postprandial insulin to help prevent meal-based excursions. The ALPHA algorithm can be used with any type of control algorithm, not just exMPC. It only uses glycemic excursion information from prior meal events to adjust the postprandial insulin delivery using an adaptive aggressiveness factor. We used 99 statistically sampled virtual patients from the OHSU virtual patient population [5] to represent patients with T1D (ie, the plant). The virtual patient population was generated based on a glucoregulatory model consisting of insulin kinetics and dynamics models and a glucose kinetics model. The parameters of the insulin dynamics model were statistically sampled to build a virtual population with different insulin sensitivities. As provided in Resalat et al. [5], the average total daily insulin requirement for these virtual patients of 44.3 units per day closely matched a clinical data set that was used to generate the virtual patients. In general, the complexity of the process model should be less than the plant for better representing real-world control scenarios whereby the human is substantially more complex than the exMPC process model. The plant in the OHSU virtual patient population is represented by eight differential equations, making it more complex than the exMPC process model.

The process model of the exMPC consists of an insulin kinetics model [6], an insulin dynamics model [7], and a glucose kinetics model as described in Resalat et al. It is presented with the following equations:

Equation 1: Process Model of the exMPC

$$
Q_{1}(t) = -\left(R + k_{21} + X(t)\right)Q_{1}(t) + k_{12}Q_{2} + PQ_{1b} + U_{G}(t)
$$
\n
$$
Q_{2}(t) = k_{21}Q_{1}(t) - k_{12}Q_{2}
$$
\n
$$
\dot{X}(t) = -P_{2}X(t) + P_{3}(I(t) - I_{b})
$$
\n
$$
X_{1}(t) = -k_{a}X_{1}(t) + u_{1}(t - \tau)
$$
\n
$$
\dot{I}(t) = -k_{e}I(t) + \frac{k_{a}}{V_{d}}X_{1}(t)
$$
\n
$$
U_{G} = \frac{D_{G}A_{G}(t - t_{0})e^{-\frac{t - t_{0}}{I_{\text{max,G}}}}}{t_{\text{max,G}}^{2}}
$$

where Q_1 and Q_2 are glucose masses in accessible and nonaccessible compartments, respectively (mg/kg). The Q_1 compartment is the glucose in plasma where it can be measured (i.e. accessible), while the *Q*² compartment is the glucose in tissue where it cannot be measured (i.e. nonaccessible). P_1 is the glucose effectiveness (min⁻¹). *X* is the effect of insulin on blood glucose (min−1). *P*² represents the decay rate of *X* (min−1). *P*³ shows the effective rate of insulin in plasma (min−2 per mU/L). The ratio between *P*³ (min−2 per mU/L) and *P*² (min−1) represents the ISF [5]. *Q*1b and *I*^b are basal plasma glucose and insulin, respectively. *k*¹² and *k*²¹ (min−1) are rate parameters describing glucose exchange kinetics, respectively. *X*I(*t*) is

the amount of insulin in the subcutaneous depot (mU/kg) , $I(t)$ is the plasma insulin concentration (mU/L) , and $u(t)$ is subcutaneous infused insulin (mU/kg/min). *k*^e is the elimination rate of insulin (min−1), *k*^a is the absorption rate of insulin (min^{-1}) , V_d is the insulin volume of distribution (L/kg), and τ is the time delay for injected insulin to be effective in the interstitial fluid (min), which was set to zero. U_G represents the glucose absorption rate from meals (mg/kg/min). $t_{\text{max,G}}$ (min) is the time-to-maximum appearance rate of glucose in Q_1 , A_G is the carbohydrate bioavailability (unitless), t_0 is the meal announcement time (min), and D_G is the estimated carbohydrate intake (mg/kg). D_G is converted from grams to milligrams per kilograms to be compatible with Q_1 in the glucose kinetics model.

The process model determines the predicted glucose levels over the prediction horizon (N_P) which are compared with the reference trajectory. See Supplementary Figure 1 and 2 below to see an example of how the exMPC model predicts glucose across the 300-minute prediction window.

Supplementary Figure 1: Example of exMPC predicting glucose

[Figure 2](#page-3-0)**:** Model estimation of glucose over the 300-minute prediction horizon within the exMPC compared with the actual glucose from the plant for patient 1 and meal scenario 1.

For glucose levels greater than the target value $(G_t = 115 \text{ mg/dL})$, the reference glucose trajectory linearly approaches the target, while for glucose levels less than the target, they exponentially approach the target as shown in Figure 1 in Resalat et al. [4].

The time constant of the exponential term is comparably low to shut off insulin faster for low glucose levels. The constraint of basal delivery was set to an upper limit of 4 times the basal rate.

1.1.1 Integrating Exercise into the Glucoregulatory Model

We used a model presented by Hernandez-Ordonez et al. [8]. The insulin sensitivity factor in [Equation 1](#page-2-2) is represented by the factors P_3/P_2 . We model the impact of exercise as an increase in insulin sensitivity according to [Equation 2](#page-3-2) whereby P_3 in [Equation 1](#page-2-2) is replaced by P_{3Ex} .

Equation 2: Impact of Exercise on Insulin Sensitivity

$$
P_{3Ex} = M_{PGU} \times M_{PIU} \times P_3
$$

 M_{PGU} represents a gain factor related to peripheral glucose uptake while M_{PGU} represents a gain factor related to peripheral insulin uptake. These parameters are a function of both percent active muscle mass (PAMM) during exercise as well as the percent of maximal $VO₂ (PVO_{2max})$ during exercise.

Equation 3: Impact of Percent Active Muscle Mass on Glucose and Insulin Uptake

$$
M_{PGU} = 1 + \frac{\Gamma_{PGUA} \times PAMM}{35}
$$

\n
$$
M_{PIU} = 1 + 2.4 \times PAMM
$$

The Γ_{PGLIA} represents the percent glucose uptake by active muscle mass and is represented by a differential equation.

Equation 4: Peripheral Glucose Uptake State Equation

$$
\Gamma_{\text{PGUA}} = -\frac{1}{30} \Gamma_{\text{PGUA}} + \frac{1}{30} \Gamma_{\text{PGUA}}
$$

The $\Gamma_{\overline{PGUA}}$ represents the peripheral glucose uptake by active tissue in steady state and is a function of PVO_{2max}, defined in [Equation 5.](#page-4-1) The PVO_{2max} is the percentage of the maximum oxygen consumption during exercise and is a function of metabolic expenditure (MET), defined in [Equation 6.](#page-4-2)

Equation 5: Impact of $PVO₂$ Max on Peripheral Glucose Uptake

$$
\Gamma_{\overline{PGUA}} = 0.006 \, (\text{PVO}_{2\,\text{max}})^2 + 1.2264 \, (\text{PVO}_{2\,\text{max}}) - 10.1958
$$

Equation 6: Estimating PVO₂ from MET

$$
PVO_{2\text{max}} = \frac{\text{MET}}{\text{MET}_{\text{max}}}
$$

MET can be derived using heart rate and accelerometry data as given in [9]. In our studies, we collect this data using a MET can be derived using a Polar M600. The Polar M600 is a smart-watch and includes a sensor for measuring heart rate and a 3-axis accelerometer. The heart rate measurement from the M600 is averaged over a 1-minute period. And the magnitude of the 3-axis accelerometry data is also averaged over a 1-minute period. Metabolic expenditure is calculated from the heart rate and accelerometry data usin[g Equation 38](#page-7-0) and further explained in section Section 6.10.

1.1.2 Mathematical Representation of the exMPC

The state space representation of the exMPC process model is represented in [Equation 7](#page-4-4) [10].

Equation 7: General Model of States in exMPC

$$
x_m(k + 1) = A_m x_m(k) + B_m u(k) + d_m(k)
$$

y(k) = C_mx_m(k)

 $x_m(k)$ is the state vector, $u(k)$ is the input vector (insulin) and $d(k)$ is the constant terms originating from the linearization of the nonlinear interactions between X and Q_1 i[n Equation 1](#page-2-2) using the Taylor's series expansion. exMPC predicts future glucose levels (y(k)) during N_P by varying future insulin levels (u(k) in (2)) during the control horizon (N_c). We defined a new vector, $x(k) = [\Delta x_m(k)^T y(k)]^T$, and re-arranged the state equations shown below. The Δ symbol represents the changes in value of a variable between k and k+1; e.g. $\Delta x_m(k+1) = x_m(k+1) - x_m(k)$.

Equation 8: States Updating in exMPC at Each Time Step

$$
\begin{bmatrix} \Delta x_m(k+1) \\ y(k+1) \end{bmatrix} = \begin{bmatrix} A_m & 0_m^T \\ C_m A_m & 1 \end{bmatrix} \begin{bmatrix} \Delta x_m(k) \\ y(k) \end{bmatrix} + \begin{bmatrix} B_m \\ C_m B_m \end{bmatrix} \Delta u(k) + \begin{bmatrix} 1 \\ C_m \end{bmatrix} \Delta d_m(k)
$$

 $y(k) = \begin{bmatrix} 0_m & 1 \end{bmatrix} \begin{bmatrix} \Delta x_m(k) \\ y(k) \end{bmatrix}$

Note that the equations above can be simplified with matrices A, B, C, and D as follows

$$
=>\begin{cases}x(k+1) = Ax(k) + B\Delta u(k) + D\Delta d_m(k) \\ y(k) = Cx(k)\end{cases}
$$

Where the state variable $x(k)$ is defined as:

 $x(k) = \begin{bmatrix} \Delta x_{\rm m}(k) \\ y(k) \end{bmatrix}$

 $\mathbf{r} = \mathbf{r} \mathbf{r}$

where, 0_m denotes a vector of zeros whose dimension is the number of the states. The predicted outputs (Y_p) are then calculated i[n Equation 9.](#page-4-6)

Equation 9: Predicted Glucose State Equation

 $Y_P = FX(k) + \Phi \Delta U + \Psi \Delta D$

where ΔU and ΔD denote the changes of the input vector, and the constant terms, respectively. The matrices F, Φ and Ψ,

which are related to A, B, C and D matrices in [Equation 8.](#page-4-5)

Equation 10: Estimation of Future Changes in Insulin Doses Across Prediction Horizon

 $Y_P = [y(k + 1) y(k + 2) ... y(k + N_P)]^T$, $\Delta U = [\Delta u(k) \ \Delta u(k + 1) ... \ \Delta u(k + N_c - 1)]^{T}$, $\Delta D = [\Delta d_m(k) \ \Delta d_m(k+1) \dots \Delta d_m(k+N_c-1)]^T$

The prediction horizon was set to 300 minutes ($N_P = \frac{300}{T_s} = 60$ samples; Ts = 5 min was the sampling interval) because

the peak effect of the short-acting insulin is several hours [11] and the control horizon was set to 20 minutes ($N_c = \frac{20}{T_s}$

4 samples) [1] . We empirically found that longer control horizons did not impact the glycemic outcomes and we chose 20 minutes to have less computational burden and faster exMPC control. The cost function, defined i[n Equation 11,](#page-5-1) consists of an error term and an input-controllable term. The error term measures the discrepancy between the predicted glucose output (Y_P) and the reference glucose trajectory (R_s) . The input-controllable term restricts the size of the input changes by a tuning parameter ($R_w = r_w \times I_{N_c \times N_c}$, where I is the identity matrix).

Equation 11: Cost Function for Optimizing Insulin

 $J = (R_s - Y_p)^T (R_s - Y_p) + \Delta U^T R_W \Delta U$

The exMPC optimization's step was done by minimizing the derivative of the cost function with respect to the input constraints shown in [Equation 12.](#page-5-2) We imposed constraints on the maximum and minimum insulin delivery rate such that the system could not deliver more than 80 units/hour or less than 0 units / hour.

Equation 12: Derivative of Cost Function Set to Zero to Optimze Insulin Dosing.

$$
\begin{cases}\n\frac{\partial J}{\partial \Delta U} = 0 \Rightarrow \Delta U = \left(\Phi^T \Phi + \overline{R_w}\right)^{-1} \Phi^T (R_s - F_X(k) - \Psi \Delta D) \\
0 < u_I < 80 \frac{\text{units}}{\text{hr}}\n\end{cases}
$$

 Γ The optimal ΔU vector contains Δu(k), Δu(k + 1),..., Δu(k + N_c – 1). However only the first element (i.e. Δu(k)), according to the receding horizon control principle, is given to the plant. Therefore, next insulin delivery shown in [Equation 13](#page-5-3) is calculated below:

Equation 13: Calculation of Optimal Insulin at Time Step k.

$$
u(k+1) = u(k) + \overbrace{[1 \ 0 \dots \ 0]}^{N_c} \Delta U
$$

We employed an observer using a Kalman filter which is shown i[n Equation 14.](#page-5-4) It compensates for the difference between the plant and the model outputs. The meal period was defined as a 4-hour period following a meal event and the non-meal period was defined as a period starting 4 hours after the meal event up to the next known meal. The non-meal period included both overnight periods and the periods between meals excluding the meal periods.

Equation 14: State Update at Time Step k.

$$
\begin{cases} x_m(k+1) = A_m x(k) + B_m u(k) + d_m(k) + K_{ob}(G(k) - C_m x_m(k)) \\ y(k) = C_m x_m(k) \end{cases}
$$

where, G is the CGM glucose level and K_{ob} is the Kalman gain. For calculating K_{ob} , we assumed that the mean and the covariance of the discrepancy between the plant's and model's outputs (model-plant mismatch) were 0 mg/dl and $100²$ $(mg/dl)^2$, respectively. In addition, we set the mean and the covariance of the process noise to 0 mg/dl and 1 $(mg/dl)^2$. We also assumed that there was no correlation between the process noise and the model-plant mismatch. To quantify the modelplant mismatch, average root mean square error over non-meal periods was used.

Equation 15: Root Mean Squared Error in Prediction.

RMSE =
$$
\frac{1}{P} \times \sum_{i=1}^{P} \sqrt{\frac{1}{N}} \times \sum_{k=1}^{N} (G(k) - y(k))^2
$$
, i = 1, 2, ...

where, N is the number of samples in a non-meal period and P is the total number of non-meal periods in a glucose signal

for one virtual patient.

An example of the accuracy of the exMPC model estimation of glucose over the 300-minute horizon compared with the actual plant glucose is given in [Figure 2](#page-3-0)**.**

2 DETAILED DESCRIPTION OF THE PREDICTIVE LOW GLUCOSE SUSPEND ALGORITHM

The exMPC control algorithm includes a predictive low glucose suspend algorithm (PLGS). The PLGS algorithm includes (1) an algorithm to predict hypoglycemia and (2) a mechanism to override the controller and shut off insulin if low glucose is predicted by the hypoglycemia prediction algorithm. The predictive algorithm is a long short term memory (LSTM) neural network and it is discussed further in Mosquera-Lopez and Jacobs [12]. In this implementation, insulin is always shut off if glucose is predicted to be less than 90 mg/dL. We have used this PLGS algorithm with a simple linear prediction model in conjunction with the a single and dual-hormone control algorithms in the past as described in [13-15]. In this study, we use the same PLGS algorithm as used in past studies, however, an LSTM model was used for prediction of low glucose. The simple linear prediction model will be used if insufficient input data is available for the LSTM to be used (i.e. 3 hours of CGM history).

3 WRIST-BASED METABOLIC EQUIVALENT TO TASK (WMETS) ALGORITHM

3.1.1 General Description

The wMETs algorithm estimates a user's real-time energy expenditure using heart rate and wrist-worn accelerometry. Energy expenditure values are expressed in metabolic equivalents (METs) that are normalized relative to a user's basal metabolic rate. The wMETs algorithm is a linear function that takes heart rate (in beats per minute, bpm) and 3-axis accelerometer readings (in g's) recorded over the past 5 minutes as inputs. The algorithm is specific to wrist-worn accelerometer devices and can estimate energy expenditure during exercise or miscellaneous activities of daily life (ADLs).

3.1.2 Algorithm Training

The dataset used to fit the linear model in the wMETs algorithm was composed of activity data from nine study participants, 21-45 years of age. After an initial screening and baseline exam, study participants were asked to perform a series of ADLs as well as three 15-minute treadmill exercise sessions of increasing intensity with at least 10-minutes of rest between sessions. Activities of daily living included 5-15 minutes each of sitting on a chair or lying on a bed, washing dishes or folding laundry, sweeping the floor or vacuuming, organizing or adjusting furniture in a room, scrubbing the carpet or the walls, and going up and down flights of stairs. The participants simultaneously wore a wrist-mounted Actigraph GT3X accelerometer, a chest-mounted Zephyr Biopatch harness, electrocardiogram (ECG) leads, and a portable VO₂ mask that collected data throughout the experiment.

The combined heart rate, accelerometry, and VO₂ data from all nine participants' exercise and ADL sessions were pooled. Roughly 60 minutes of ADLs and 120 minutes of exercise data for each participant were included in the final dataset. All VO2 data for each participant were binned into 1-minute segments, averaged, and used as ground-truth metabolic expenditure to fit a linear model. A linear regression model was fit using the ordinary least squares method. All data preparation, model fitting, and analysis were completed using Python (version 3.7.6) and scikit-learn (version 0.22.1).

3.1.3 Equation

At the time of prediction, all heart rate data from the prior 5-minutes for a given participant are binned into (5) 1-minute segments and averaged (HRlag). An activity metric, Euclidean norm minus one (ENMO), is calculated from each raw, 3axis accelerometer vector according to [Equation 37:](#page-6-5)

Equation 16: Euclidean Norm Minus One

 $ENMO_t = \left| \sqrt{x_t^2 + y_t^2 + z_t^2} - 1 \right|$

Where x_t , y_t , and z_t are the raw acceleration values of the x, y, and z-axes (in g's), respectively, at time t.

All ENMO_t values for a given participant are calculated over the prior 5-minutes, binned into (5) 1-minute segments and averaged (ENMO_{lag}). Metabolic equivalents (METs) are estimated according to [Equation 38](#page-7-0) using the ten minute-level averaged HRlag and ENMOlag values:

Equation 17: wMETs Algorithm for Metabolic Expenditure Estimation

```
wMETs = -3.282 \cdot 10^{-3} * HR_{laqA} +
         -6.261 \cdot 10^{-3} * HR_{laq3} +
        1.021 \cdot 10^{-2} * HR_{laq2} +
        5.990 \cdot 10^{-3} * HR_{laq1} +
        -1.615 \cdot 10^{-3} * HR_{laq0} +
        -5.667 \cdot 10^{-1} * ENMO_{laq4} +
        2.597 * ENMO_{lag3} +4.100 * ENMO_{laq2} +6.252 * ENMO_{1 a a 1} +-2.131 * ENMO_{laq0} +1.115
```
Where:

- HR_{lago} is the average of all heart rate samples (bpm) received between the current time and one-minute prior,
- HR_{lag1} is the average of all heart rate samples (bpm) received between one-minute and two-minutes prior to the current time,
- HRlag2 is the average of all heart rate samples (bpm) received between two-minutes and three-minutes prior to the current time,
- HRlag3 is the average of all heart rate samples (bpm) received between three-minutes and four-minutes prior to the current time, and
- HR_{lag4} is the average of all heart rate samples (bpm) received between four-minutes and five-minutes prior to the current time.
- $EMMO_{lae0}$ is the average of all ENMO values calculated from accelerometer samples received between the current time and one-minute prior,
- ENMO $_{\text{lag1}}$ is the average of all ENMO values calculated from accelerometer samples received between oneminute and two-minutes prior to the current time,
- ENMO $_{\text{lag2}}$ is the average of all ENMO values calculated from accelerometer samples received between twominutes and three-minutes prior to the current time,
- $EMMO_{lag3}$ is the average of all ENMO values calculated from accelerometer samples received between threeminutes and four-minutes prior to the current time, and
- ENMO $_{\text{large}}$ is the average of all ENMO values calculated from accelerometer samples received between fourminutes and five-minutes prior to the current time.

4 PROTOCOL

PROTOCOL TITLE: A randomized, two-way, cross-over study to assess the efficacy of an exMPC exerciseenabled closed-loop system vs exAPD exercise-enabled closed-loop system

STUDY SITE: Oregon Health Science University

Background:

Closed-loop systems automate insulin delivery based on continuous glucose monitoring (CGM) values to minimize hypoglycemia and hyperglycemia, and in some cases also deliver glucagon to prevent and treat hypoglycemia. Closedloop systems have been tested extensively in both the inpatient and outpatient settings. Our group has previously developed both an insulin-only closed-loop system and a dual-hormone (insulin and glucagon) closed-loop system. The novelty of these closed-loop systems are their ability to automatically detect exercise and adjust dosing in response to the exercise. Recently, we completed an outpatient closed-loop study, which tested the OHSU insulin-only closed-loop system with automated exercise detection [16]. Participants participated in four arms in randomized order: current care (their typical diabetes care, which included an insulin pump with or without a sensor), sensor-augmented pump with predictive low glucose suspend (using a t:slim pump and Dexcom G5 sensor with a PLGS algorithm running on the smartphone), insulin-only closed-loop system with exercise detection, and bi-hormonal (insulin + glucagon) closed-loop system with exercise detection. Each arm was 4 days long, with two 12 hour in clinic visits on days 1 and 4, with the remainder of the time spent as an outpatient using a cloud-based remote monitoring system. During the time spent as an outpatient, participants went to work, slept at home, travelled, and even went skiing. Participants performed moderate exercise for 45 minutes at 60% VO2max during both of the in clinic visits as well as performed one at home exercise session. The primary endpoints were percent time in hypoglycemia (CGM <70 mg/dL) and percent time in target (CGM 70-180 mg/dL) expressed as mean (SD). The mean time in hypoglycemia was the lowest with dual-hormone during the exercise period: $3.4\pm4.5\%$ vs. $8.3\pm12.6\%$ single-hormone (p=0.009) vs 7.6 $\pm8.0\%$ predictive low glucose suspend $(p<0.001)$ vs $4.3\pm6.8\%$ current care where pre-exercise insulin adjustments were allowed (p=0.49). This type of manual adjustment was not allowed in any other arm and no snacks were given prior to exercise. Time in hypoglycemia was also the lowest with dual-hormone during the entire 4-day study: $1.3\pm1.0\%$ vs. $2.8\pm1.7\%$ single-hormone (p<0.001) vs. 2.0 \pm 1.5% predictive low glucose suspend (p=0.04) vs. 3.1 \pm 3.2% current care (p=0.007). Time in range during the entire study was the highest with insulin-only closed-loop system: 74.3±8.0% vs 72.0±10.8% dual-hormone (p=0.44). The OHSU insulin-only closed-loop system performance was comparable to those reported in a recent meta-analysis of AP studies [17] which reported a mean time in range of 70.8% using insulin-only closed-loop systems vs 74.3% in our recent study. We have also completed a study on patients with T1D who used the OHSU dual-hormone AP for 22 hours within the hospital during which participants performed mild exercise for 45 minutes on a treadmill [15]. For this study, we used the OHSU bi-hormonal artificial pancreas system that adjusts dosing after an exercise announcement to reduce exerciserelated hypoglycemia. Results showed that adjusting hormones during exercise reduced hypoglycemia.

The study described within this protocol is designed to test the efficacy of a new modified insulin-only closed-loop algorithm, a model-predictive control (exMPC) algorithm that modulates insulin delivery based on estimated activity level. The potential benefit of this type of algorithm is that it handles exercise not as a discrete event, but it automatically adjusts insulin delivery based on estimated activity level. This type of algorithm may significantly improve glucose control over the exAPD algorithm which is designed only to detect exercise over 4 metabolic equivalent of tasks (METs) for a specific duration of 45 minutes.

Primary Objective:

• To confirm superiority of the OHSU exMPC exercise-enabled closed-loop system as measured by percent of time with sensed glucose less than 70 mg/dl as compared to the OHSU exAPD exercise-enabled closed-loop system.

Secondary Objective:

• To confirm superiority of the OHSU exMPC exercise-enabled closed-loop as measured by other glucose metrics as compared to the OHSU exAPD exercise-enabled closed loop system.

Study Hypothesis:

We propose that the use of the OHSU exMPC exercise-enabled closed-loop system as compared to the OHSU exAPD exercise-enabled closed-loop system will increase the time in range as measured by sensed glucose values.

Endpoints

Primary Endpoint:

• Percent of time with sensed glucose <70 mg/dl across the duration of inpatient day (Day 3 exMPC vs. Day 1 exAPD)

Secondary Endpoints:

- Percent of time with sensed glucose between 70-180 mg/dl across the study duration
- Percent of time with sensed glucose between 70-180 mg/dl across the duration of the inpatient day (Day 3 exMPC) vs. Day 1 exAPD)
- Percent of time with sensed glucose between $70 180$ mg/dl from the start of the in-clinic exercise session until the start of the next meal
- Percent of time with sensed glucose \leq 90 mg/dl from the start of the in-clinic exercise session until the start of the next meal
- Number of carbohydrate treatments (defined as 15 or 20 grams of carbohydrate)
- Mean sensed glucose
- Coefficient of variation of glucose
- Percent of time with sensed glucose <54 mg/dl
- Percent of time with sensed glucose >180 mg/dl
- Mean amount of insulin delivered per day (in units/kg)

Study Type

This is a single center, randomized, two treatment, crossover trial designed to compare the glucose control resulting from the use of the OHSU exMPC exercise-enabled AP system as compared to the OHSU exAPD exercise-enabled AP system.

Study Population

Study population will be adults with type 1 diabetes, ages $21 - 50$ years of age. Older participants are excluded due to higher risk of unrecognized coronary artery disease. Younger participants are excluded as it is appropriate to assess safety first in the adult population. Twenty-five participants will be recruited to participate in studies.

Power Analysis

A sample of 24 subjects provides at least 80% power to detect a paired difference for a two-sided Wilcoxon signed-rank test of the difference between use of exercise-enabled exMPC algorithm and exAPD algorithm with alpha set at 0.05 for our primary outcome of percent of time <70 mg/dl. We use the sample size for a one-sample t-test, n, with an adjustment factor for the double exponential (or Laplace) distribution of the difference such that the final sample size $n' = n/(2/3)$. We anticipate a mean difference of 1.2 (SD 2.5) based on our prior published AP study data. For normally-distributed secondary outcomes, we have >80% power to detect differences of 0.6 SD or greater using a two-sided one-sample t-test for the mean difference at the 0.05 significance level. These calculations were performed using PASS version. We propose to enroll 25 participants in this study as this will yield more than 80% power as stated above (see Participant Recruiting).

Protocol Summary:

Participants will undergo two approximately 76 hour studies. See **Figure 1 &2** for a diagram of the study flow and structure. During each of these intervention visits, participants will wear an Omnipod to deliver insulin and a Dexcom G6 CGM to measure glucose. The CGM system will provide sensed glucose data every 5 minutes. Sensed glucose data will be wirelessly transmitted via Bluetooth Low Energy (BTLE) from the Dexcom G6 transmitter to the smartphone master controller every five minutes. The smart phone will wirelessly communicate via BTLE to an Omnipod through a PDM (Insulet Corp.). During one of the studies, glucose will be controlled using the exMPC exercise-enabled mode. During another study, glucose will be controlled using the exAPD exercise-enabled mode. The closed loop system will receive activity data through a Polar M600 watch worn by the participant. Participants will arrive at the clinic at approximately 7am for the inpatient visits. Participants will eat breakfast, lunch and dinner at approximately 8am, noon, and 5pm respectively. Participants will perform activities of daily living and exercise. Participants will be discharged at approximately 7pm. Participants will then go home for the remainder of the study visit and perform two exercise sessions at home. Participants will return to OHSU on Day 4 for removal of all devices. The exception to this is for the first 8 participants using the exMPC exercise-enable mode. These participants will stay at OHSU during the day and go home to sleep each night (7pm-7am). These participants will use the system in open-loop mode while off campus.

Figure 1: Study Flow Design

Figure 2: Study structure

Participant Criteria

Inclusion Criteria:

- 1. Diagnosis of type 1 diabetes mellitus for at least 1 year.
- 2. Male or female participants 21 to 50 years of age.
- 3. Physically willing and able to perform aerobic exercise (as determined by the investigator after reviewing the participant's activity level)
- 4. Current use of an insulin pump for at least 3 months with stable insulin pump settings for ≥ 2 weeks.
- 5. Lives with another person age 18 or older who will be present while participant exercises at home and that can attend the training on using the system.
- 6. Lives within 40 miles of OHSU main campus.
- 7. HbA1c $\leq 10\%$ at screening.
- 8. Total daily insulin requirement is less than 139 units/day.
- 9. Current use of a phone or other device so can be contacted by study staff off-campus
- 10. Willingness to follow all study procedures, including attending all clinic visits.
- 11. Willingness to sign informed consent and HIPAA documents.

Exclusion Criteria:

- 1. Female of childbearing potential who is pregnant or intending to become pregnant or breast-feeding, or is not using adequate contraceptive methods. Acceptable contraception includes birth control pill / patch / vaginal ring, Depo-Provera, Norplant, an IUD, the double barrier method (the woman uses a diaphragm and spermicide and the man uses a condom), or abstinence.
- 2. Any cardiovascular disease, defined as a clinically significant EKG abnormality at the time of screening or any history of: stroke, heart failure, myocardial infarction, angina pectoris, or coronary arterial bypass graft or angioplasty. Diagnosis of $2nd$ or $3rd$ degree heart block or any non-physiological arrhythmia judged by the investigator to be exclusionary.
- 3. Renal insufficiency (GFR < 60 ml/min, using the MDRD equation as reported by the OHSU laboratory).
- 4. Liver failure, cirrhosis, or any other liver disease that compromises liver function as determined by the investigator.
- 5. Hematocrit of less than 36% for men, less than 32% for women.
- 6. Hypertensive participants with systolic blood pressure ≥ 160 mmHg or diastolic blood pressure ≥ 100 mmHg despite treatment or who have treatment-refractory hypertension (e.g. requiring four or more medications).
- 7. History of severe hypoglycemia during the past 12 months prior to screening visit or hypoglycemia unawareness as judged by the investigator. Participants will complete a hypoglycemia awareness questionnaire. Participants will be excluded for four or more R responses.
- 8. History of diabetes ketoacidosis during the prior 6 months prior to screening visit, as diagnosed on hospital admission or as judged by the investigator.
- 9. Adrenal insufficiency.
- 10. Any active infection.
- 11. Known or suspected abuse of alcohol, narcotics, or illicit drugs.
- 12. Seizure disorder.
- 13. Active foot ulceration.
- 14. Severe peripheral arterial disease characterized by ischemic rest pain or severe claudication.
- 15. Major surgical operation within 30 days prior to screening.
- 16. Use of an investigational drug within 30 days prior to screening.
- 17. Chronic usage of any immunosuppressive medication (such as cyclosporine, azathioprine, sirolimus, or tacrolimus).
- 18. Bleeding disorder, treatment with warfarin, or platelet count below 50,000.
- 19. Allergy to aspart or lispro insulin.
- 20. Current administration of oral or parenteral corticosteroids.
- 21. Any life threatening disease, including malignant neoplasms and medical history of malignant neoplasms within the past 5 years prior to screening (except basal and squamous cell skin cancer).
- 22. Beta blockers or non-dihydropyridine calcium channel blockers.
- 23. Current use of any medication intended to lower glucose other than insulin (ex. use of liraglutide).
- 24. A positive response to any of the questions from the Physical Activity Readiness Questionnaire with one exception: participant will not be excluded if he/she takes a single blood pressure medication that doesn't impact heart rate and blood pressure is controlled on the medication (blood pressure is less than 140/90 mmHg).
- 25. Any chest discomfort with physical activity, including pain or pressure, or other types of discomfort.
- 26. Any clinically significant disease or disorder which in the opinion of the Investigator may jeopardize the participant's safety or compliance with the protocol.

Participant Recruiting:

Participants will be recruited from OHSU clinics, from flyers to be posted in approved places at OHSU or posted on the web to the clinical trials page for the OHSU Schnitzer Diabetes Clinic, to the Clinic's facebook group, ads on facebook, electronic newsletter or from the OHSU Participant Recruitment website. The T1D Exchange may send out approved recruitment emails to Glu community users in the Portland/Vancouver area. Handouts may also be made available to faculty at Tuality, Providence, Kaiser, and Legacy to pass along to patients/participants who show interest in the study. Records from OHSU Schnitzer Diabetes Clinic patients may be screened to find potential participants. Participants will also be recruited from a list of participants who participated in past OHSU studies who have agreed to be contacted regarding future studies involving Drs. Castle from the OHSU diabetes research registry and/or [www.clinicaltrials.gov.](http://www.clinicaltrials.gov/) Non-english speaking participants will not be recruited since this protocol would require the use of medical devices and mobile software that do not have non-english versions available. Up to 50 participants may be screened in this study. Goal enrollment is 25 participants.

Withdrawal Criteria

The participant may withdraw at will at any time or at the discretion of the Investigator.

A participant must be withdrawn if the following applies:

- Hypoglycemia during the treatment period posing a safety problem as judged by the investigator.
- Hyperglycemia during the treatment period posing a safety problem as judged by the investigator.
- Protocol deviation having influence on efficacy or safety data as judged by the Investigator.
- Substantial and repeated non-compliance with trial procedures.
- Pregnancy.
- Intention of becoming pregnant.

Visit Procedures

Screening (Visit 1)

Screening will take place within 12 weeks prior to the 1 week run-in period (Visit 2). All screening visits will take place at OHSU's Oregon Clinical Translational Research Institute (OCTRI) outpatient clinic, the AIMS lab in Biomedical Engineering Dept at CHH1 or the Harold Schnitzer Diabetes Health Center. The participant will be sent the consent form prior to the screening by email so that they can have time to read it fully at their leisure and prepare any questions they might have. Upon arrival and prior to any procedures, study staff will explain the study, give the participant ample time to ask questions and consider participation, and ensure that the participant voices understanding of the informed consent and study requirements. To minimize the possibility of coercion and to ensure that participant is signing the appropriate version of the consent, an informed consent checklist will be used by study staff. After the participant has signed the consent, a copy of the consent/authorization form will be given to the participant. The original will be kept for the source document.

A capillary blood glucose (CBG) will be obtained and measured by a Contour Next glucose meter and recorded after consenting. Prior to measurement of any blood samples, the meter will undergo quality control testing with two different glucose levels, one high and one low, and both values must fall within the accepted range for a meter to be used.

Study personnel will review medical history, and medications. Height, weight, pulse, and blood pressure will be obtained. A study investigator will perform a physical examination, excluding breast and pelvic exams. Females of child-bearing potential will take a urine pregnancy test, which must be negative to participate. A venous blood sample will be taken for the following tests: hemoglobin A1C, complete blood count, complete metabolic set (including creatinine, liver set, and electrolytes). If the participant has already had these lab studies done within the past 2 months and the results are available, those results can be used in place of the blood draw. An EKG will be completed. A study investigator will assess inclusion/exclusion criteria and review the participant's medical record for clarification as needed. When feasible, the participant's insulin pump and glucose sensor, if applicable, will be downloaded before enrollment. A three-digit participant ID number will be assigned to the participant. This visit will take approximately 1.5 hours.

One Week Run-in Period

The purpose of this run-in period is to teach participants how to use the Dexcom CGM system using the iPancreas software. For those new to CGM, it is also designed to provide them exposure prior to starting the intervention visits. The one week run-in period will take place within 8 weeks prior to the first 76 hour treatment visit. After arrival at the OHSU OCTRI outpatient clinic, the AIMS lab in Biomedical Engineering Dept at CHH1 or Harold Schnitzer Diabetes Health Center clinic, women of childbearing potential will receive a urine pregnancy test if the last pregnancy test was more than 7 days prior. This test must be negative before further participation is allowed. This visit will take approximately 30-60 minutes.

Participants will receive training on how to use and calibrate the Dexcom G6 CGM system including changing out the sensor every 10 days. The wire glucose sensor is sterile and commercially available from DexcomTM and will be used for single use only as directed by the manufacturer. Participants will be trained to insert the sensor into the subcutaneous tissue of the abdomen after appropriate preparation of the abdominal skin per the manufacturer's directions. Participants will be trained how to pair the Dexcom G6 transmitter to the iPancreas app on the smart phone, start and stop a new sensor session and how to enter calibrations. The Dexcom G6 does not require calibrations. As part of the training, study staff will review with the participants that Dexcom G6 values can be inaccurate. In the event that the participant's symptoms (such as symptoms of hypoglycemia or hyperglycemia) are discrepant with the G6 CGM reading, then the participants will be instructed to perform a CBG and use this CBG value to make treatment decisions and use the CBG value to calibrate the Dexcom G6 device. Participants will be provided with a copy of the Dexcom G6 user guide.

The CGM alerts will be set at 70 mg/dL and 300 mg/dL. Participants will be trained how to begin a run-in study on the iPancreas app at home after two hours when the sensor warm-up is completed. Participants will be given a Dexcom G6 transmitter and sensor to insert the day before each treatment visit along with a Contour Next meter for measuring their capillary blood glucose.

There is the option to complete this visit via Webex with the study devices delivered to the participant and study staff virtually connecting with participants for training on the devices and study procedures while they are at home. Participants will complete a urine pregnancy test at home provided by study staff if the last pregnancy test was more than 7 days prior.

76 hour Treatment Visits

The participant will be asked to check his/her glucose before driving to the clinic and to bring a snack in the car in case hypoglycemia does occur (in which case, the participant must park and treat the hypoglycemia). After the first treatment visit, the washout period will be 6 days to 10 weeks calculated from the day of admission to the research center until the start of the next admission. The participant will arrive at the research center at approximately 7am. Women of childbearing potential will receive a urine pregnancy test if the last pregnancy test was more than 7 days prior. This test must be negative before further participation is allowed

An assessment will take place at the beginning of the participant's first closed-loop study. We will inquire whether the participant has had changes in their medications and/or medical history to confirm the participant hasn't developed any study exclusion criteria. A capillary blood glucose (CBG) will be obtained and measured by a Contour Next glucose meter. When they arrive, participants will be given 15-20 grams of oral carbohydrate if the CBG reading is less than 70 mg/dl. CBG values $>$ 300 mg/dl will be managed at the discretion of the investigator with a correction bolus and serum ketones will be checked. If serum ketones are ≥ 0.6 mM, the study will not be started and insulin therapy will be guided by the onsite investigator.

During each treatment visit, glucose will be controlled using either: 1) the OHSU exMPC exercise-enabled closed-loop mode or 2) the OHSU exAPD exercise-enabled closed-loop mode. The first 12 hours of the visit will be conducted in the OCTRI inpatient research unit, the Harold Schnitzer Diabetes Health Center or the AIMS lab in the Biomedical Engineering Dept at CHH1. The participants will then go home. For the first 8 participants using the exMPC mode, they will stay at OHSU during the day and go home at night, using the system in open-loop mode while off campus. If safety criteria are met, as outlined below, subsequent participants will be at home for the remainder of the intervention period, returning on Day 4 to the clinic to return all devices. A code team is available immediately by page at all times.

Inpatient Visits (~40 hours for first 8 participants using exMPC, 12 hours for remainder of participants)

An Omnipod (510K#042792) will be filled with Novolog[®] or Humalog[®] insulin for all studies. We will use only name brand insulin, not generic insulin. We will provide the participants with kits to replace the pod at home in case of pod malfunction or dislodgement at home. The Omnipod will be primed and inserted as directed by the manufacturer. Research staff will train participants on the approved pod placement options from the Omnipod User Guide. Participants will wear a Polar watch to inform the controller. Participants will disconnect his/her own pump and remove his/her own insulin infusion set once insulin delivery has started via the Omnipod. The research staff will initialize the system and begin the closed-loop study with the correct mode, either exAPD or exMPC based on the participant's randomization, using the iPancreas smart phone application. Two staff will review each of the settings to confirm the settings are inputted correctly. The participant will be given enough study supplies for the 76 hour visit. A study investigator will be available in person or by phone for the entire 76 hour visit.

Study staff will complete a training with each participant on using the Dexcom G6, the Omnipod system, and the Polar watch. Participants will be shown how to use the smartphone user interface which includes: entering basal profiles, insulin carb ratios and sensitivity, activating and deactivating Omnipods, giving meal boluses, carb treatments, blood glucose values and ketone levels, addressing alerts, troubleshooting the devices connection to the phone via Bluetooth, and pausing the study. Instruction will be given on identifying an Omnipod malfunction. The time required for this training will vary, depending on the experience of each participant, but will be sufficient to help him/her become comfortable

using the smart phone and changing the Omnipod. If the participant experiences difficulties using the Omnipod during the study period, study staff will be available to educate and support by phone. Training will include that the smart bolus calculator available in open loop mode is a suggestion and that participants should use their judgement on what insulin dose to take as the bolus calculator is not aware of certain circumstances such as illness or alcohol use.

The participant will need to demonstrate competency in operating the system before study staff leave the room. A competency assessment will be completed throughout the day of the first closed-loop visit. Each participant will start the G6 sensor and start the Omnipod on their own during onboarding. Participants will need to show competency in accepting exercise detection, entering fingerstick glucose values and announcing meals during the day on campus. Participants will demonstrate competency in the following using a simulated closed loop study: pausing and resuming a study, using the smart bolus calculator and manual bolus while in pause mode, for exAPD: force start exercise, cancel exercise, for exMPC: disabling exercise adjustment. Participants will be given a pager number to call for any problems during the 76 hour visit. The companion will accompany the participant to receive training (or be previously trained) on treatment in case of severe hypoglycemia episode, including administration of rescue carbohydrates and use of emergency glucagon kit. Companions will also be trained on the closed-loop system. This training may be completed virtually.

The algorithm will push data up to a cloud server that can be monitored remotely every 5 minutes. A study coordinator will be available at all times for the visits with the ability to monitor iPancreas remotely via a cloud system on the web in the event of any issues. iPancreas will generate alerts on the smartphone according to Appendix D. The participants will all be trained as to the action required by each alert. Each alert has a specific condition to be met for it to be considered serviced (i.e. enter treatment with oral carbohydrates). The refractory period is also specific to each alert with shorter refractory periods for alerts that concern participant safety. If the participant does not appropriately respond to the alert in the allotted timeframe, the alert will push to the study coordinator and the on-call investigator according to Appendix D. At that time, the coordinator will pull up the web-based monitoring system. The study investigator and technician may intervene with a telephone call, text or a personal visit at any time. If the participant cannot be reached and sensor glucose is below 50 mg/dl, the emergency contacts provided by the participant will be contacted. If the alert is still un-serviced and study staff are unable to reach the participant or either of the emergency contacts, emergency medical services will be contacted. To facilitate this, the phone will track the participant's location and push GPS coordinates to the cloud server approximately every 10 minutes. Cloud coordinates will be pushed with a known, fixed offset to allow for scrambling.

In order to push alerts to study coordinators and study investigators, the cloud server used for remote monitoring will have a drop down menu for study staff to sign in and out for the duration of their monitoring shift. Each study coordinator and investigator listed in the menu will have a cell phone number on file that can receive texts with pushed alerts.

During all studies, sensed glucose data will be wirelessly transmitted via BTLE from the Dexcom G6 transmitter to the insulin-delivery algorithm every five minutes. The insulin-delivery algorithm will calculate insulin doses and will run on a smartphone. The smart phone will wirelessly communicate via BTLE to a PDM communicating to an Omnipod for automated insulin delivery.

Both the exAPD and exMPC control algorithms include a predictive low glucose suspend algorithm (PLGS). The PLGS algorithm includes (1) an algorithm to predict hypoglycemia and (2) a mechanism to override the controller and shut off insulin if low glucose is predicted by the hypoglycemia prediction algorithm. The predictive low glucose suspend algorithm will be continuously monitoring the trend of blood glucose with every new glucose value. If the sensed glucose is within 70-140 mg/dl and predicted to fall below 90 mg/dl within the next 30 minutes according to the LSTM algorithm, insulin infusion will be turned off. If this occurs, the control algorithm will shut off insulin for a maximum of 120 minutes suspension within any 150 minute window. Suspension will be limited to 180 minutes during the nighttime (11pm-7am). After 5 minutes of insulin suspension, insulin delivery will resume if sensor glucose is above 70 mg/dl and predicted to rise above 120 mg/dl at 30 minutes in the future.

If at any time study staff determines that a sensor can no longer be used, a new sensor will be inserted. In order to ensure safety and to assess sensor accuracy, the participant will be asked to check their blood glucose two times during the day (typically before breakfast and at bedtime), for symptoms of hypoglycemia, and in response to system alert (such as for

low or high sensor alerts). We will also instruct patients to perform a capillary blood glucose immediately prior to and immediately following exercise. Participants will be instructed to check a fingerstick glucose at bedtime and again at 2am on exercise days. The participant will be able to check his/her capillary blood glucose more than 2 times a day if they feel they need to. If the participant's blood glucose is \leq 70 mg/dl or is experiencing symptoms of hypoglycemia, he/she will be instructed to treat with 15 grams of carbohydrates.

For participant safety, if a sensor value is not available for 20 minutes or communication with the Omnipod is lost for more than 30 minutes, the Omnipod will begin insulin delivery according to a pre-set basal profile(s) inputted for the participant at study start. When this occurs for a lost sensor, iPancreas will activate the predictive low glucose suspend feature if the last known sensor value was within the range of 70-140 mg/dl and predicted to fall below 90 mg/dl within thirty minutes or if the sensor glucose is less than 70 mg/dl. Insulin is suspended for 30 minutes, after which basal insulin delivery resumes. When communication with the sensor or Omnipod is restored, the system will automatically resume, updating the IOB accordingly.

Participants will eat breakfast, lunch and dinner at approximately 8am, noon and 5pm respectively. Meals will be announced to the controller. For each meal, food items will be self-selected. The number of grams of carbohydrates will be counted by the participant and entered in the controller. Immediately before eating, each participant will receive 100% (nominal) of his/her typical pre-meal insulin dose based on their insulin to carbohydrate ratio. The same self-selected meals will be offered during exAPD Day 1 and exMPC Day 3 inpatient days, and in the event a particular food item is not available, a different item with a similar amount of carbohydrate will be provided.

Participants will perform activities of daily living (ADLs), such as vacuuming, washing dishes, and folding laundry for approximately 30 min at 10am or 3pm depending on the study (see 'Safety assessment – first 8 participants using exMPC'). Participants will complete a 30-minute aerobic exercise video at 10am or 3pm. (see 'Safety assessment – first 8 participants using exMPC'). See Figure 3 below.

Figure 3: Diagram of Exercise Study Procedures

See Appendix E for Aerobic Video Outline. For participant safety, capillary blood glucose must be 80 mg/dl or higher to begin exercise.

During the exercise period, there will be defined rules for stopping exercise, including:

- If the participant feels unwell,
- If the participant develops hypoglycemic symptoms, such as excessive sweating, shaking/tremors, palpitations, feelings of dread or panic, light-headedness, nausea, difficulty concentrating or the like,
- If the participant develops chest pain/pressure,
- If the participant develops undue shortness of breath (SOB),
- Signs of poor perfusion: light-headedness, confusion, ataxia, pallor, cyanosis, nausea, or cold and clammy skin
- If the maximum heart rate of the participant (MHR) is exceeded,
- For patient preference.

If the exercise is stopped prematurely, the duration of exercise will be noted by the study staff and if the participant is deemed safe to participate in future studies, the exercise will be stopped after that same time duration for subsequent studies. For participant safety, if capillary blood glucose is \leq 70 mg/dl at any point during the exercise period, the participant will consume 15 g of carbohydrates and delay completion of exercise until blood glucose raises above 80 mg/dl. If glucose fails to rise above 80 mg/dl after a carbohydrate treatment during exercise, a second treatment of 15 grams will be given. Participants will be given instructions to take additional carbs at the start of exercise as needed based on CGM and trend to avoid post-exercise lows. See Appendix F inpatient instruction card.

Participants will wear a Polar watch for collecting heart rate and accelerometry data. The Polar watch transmits this data to the smartphone controller via Bluetooth. iPancreas will convert the heart rate and accelerometry data into an estimated energy expenditure to determine if the participant is exercising. If the communication is not working between the Polar watch and the smartphone at the time of exercise during the inpatient visit, exercise may be delayed until communication is restored. We are aware that there is a risk of hyperglycemia if the participant stops exercising after a short time with continued adjustments to insulin. When using the exAPD mode: 1) an exercise cancellation option is available on the user interface for up to 30 minutes after the start of exercise that will revert insulin parameters to their nominal values and 2) if exercise is not detected by the algorithm when the participant is actually exercising, an exercise announcement button on the iPancreas user interface will be used. When using the MPD mode, the user can disable the exercise adjustment, thereby removing incoming exercise data for insulin dosing calculations.

Safety assessment – first 8 participants using exMPC

Given this is the first human testing of the exMPC exercise-enabled mode, the first 8 participants will stay at OHSU during the day for the entire period of using the exMPC algorithm and go home to sleep each night (approximately 7pm-7am), using the system in open-loop mode while off campus. These participants will still complete only one in-clinic day for the exAPD arm (Day 1).

The study schedule will be slightly different for the days spent on campus using the exMPC algorithm. On Day 1, participants will perform ADLs at 10am and again at 3pm. On Day 2, participants will complete the study aerobic exercise video at 10am and ADLs at 3pm. On Day 3, participants will perform ADLs at 10am and complete the study exercise video at 3pm. For each meal, food items will be self-selected. See Figure 3 above.

For these 8 participants, we are matching Day 1 of the exAPD arm to Day 3 of the exMPC arm. Participants will also perform the same activities at the same time on these days, ADLs at 10am and the study exercise video at 3pm.

Capillary blood glucose will be measured with two consecutive blood glucose measurements at least 30 minutes apart prior to discharge. Participants will wait to discharge home at approximately 7pm if capillary blood glucose is <85 and or >300 mg/dl or at the discretion of the study investigator, and treatment will be at the discretion of the study investigator. Investigator will review the physician portal to assess participant's insulin on board before they go home. These participants will use the system in open-loop mode while at home. Participants will be given a snack to take with them in case they develop hypoglycemia or hypoglycemic symptoms on the commute home.

We will complete an interim analysis after the first 8 subjects using the exMPC assessing time spent with CGM < 54 mg/dL , $\langle 70 \text{ mg}/dL$, and $> 300 \text{ mg}/dL$ and any adverse events. The exMPC data will be compared to the equivalent data from the exAPD studies. If the interim analysis shows that the exMPC data is comparable to the exAPD studies and presents no safety concerns, we will proceed to studying the exMPC during outpatient visits. If it does not or if there are any instances of severe hypoglycemia or DKA (unrelated to infusion set failure), then the study will be halted, changes will be made as appropriate to algorithm and the protocol and a plan will be submitted to FDA and IRB that will need to be approved before further studies are conducted. After the algorithm adjustments are made, an additional 8 participants will then stay at the research clinic during the day for the entire period of using the exMPC algorithm.

The remainder of the participants using the exMPC algorithm will be allowed to complete the study as an outpatient after the first approximately 12 hour study visit after 8 participants are deemed safe to sleep at home and successfully completed the exMPC arm without any events of severe hypoglycemia or diabetic ketoacidosis. Severe hypoglycemia is defined as event that required the assistance of another person due to altered consciousness, and required another person to actively administer carbohydrate, glucagon, or other resuscitative actions. Diabetic ketoacidosis is classified as: symptoms such as polyuria, polydipsia, nausea, or vomiting, serum ketones >1.5 mmol/L or moderate/large urine ketones, either arterial blood pH<7.3 or venous pH<7.24 or serum bicarbonate <15, and treatment provided in a health care facility. If an event of severe hypoglycemia or diabetic ketoacidosis should occur, the study will be halted and a description of the serious adverse event and a new risk mitigation plan will be submitted to the FDA. The study will only resume once FDA approval is received.

We are aware that there is a risk of severe hypoglycemia while the participants are at home. The system will alert if the sensed glucose values fall below 70 mg/dl prompting the participant to obtain a capillary blood glucose sample. Participants will be required to live with at least one other person age 18 or older and live within 40 miles of OHSU. All participants will provide two emergency contacts to study staff and will be given an emergency glucagon hypokit if they don't already have one.

In case of a system error that cannot be corrected immediately with the participant off campus, the participant will be able to go into open-loop mode. This will allow the Omnipod to begin basal insulin delivery for a pre-set basal profile(s) inputted for the participant at the study start. Participants will be able to give meal boluses and corrections through the Omnipod while in open-loop mode. When the error is resolved, the participant can resume closed-loop mode and the system will resume. If the participant goes into open-loop mode, this will be visible on the cloud server and may prompt a telephone call from study staff to determine the issue and the best way to resolve it. The participant may contact study staff at any time during the outpatient portion of the visit.

Discharge from inpatient clinic

If the participant's schedule does not allow them to return to the clinic before the 72 hour mark from the time the first pod was activated, then a new insulin pod will be placed as described above approximately 1-2 hours prior to discharge to avoid use of any single pod beyond 72 hours. At the completion of the 12 hour inpatient visit, participants will be discharged from the clinic, with the exception of the first 8 participants using exMPC exercise-enable mode, as detailed above. Capillary blood glucose will be measured prior to discharge. Participants will wait to discharge home if capillary blood glucose is <85 and or >300 mg/dl or at the discretion of the study investigator, and treatment will be at the discretion of the study investigator. Once participant's capillary blood glucose is between 85-300 mg/dl, they can be discharged home. For all participants using the exMPC mode, the investigator will review the physician portal to assess participant's insulin on board before they go home. Participants will be given enough supplies to continue running the study while at home for 2 days. After the participant is discharged on Day 1, he/she will return home for Day 2 and Day 3 and return to OHSU on the morning of Day 4 to return all devices and end the study. Participants will be instructed to not eat a meal after he/she leaves the inpatient clinic after the Day 1 visit. Starting the morning of Day 2, participants will be instructed to eat meals and snacks as they normally would. Participants will be instructed that his/her companion will be required to stay with them each night while an outpatient.

For the two days spent as an outpatient, participants will be asked to workout using the study video in the morning at a time of their choice on either Day 2 or Day 3. Participants will be asked to fill out the specifics of the exercise in a journal. Participants will be asked to not exercise on the other day spent at home. During the exAPD visit, participants will be instructed to only accept exercise detection on the app if they are exercising for 20 minutes or longer. Participants will be reminded to check their blood glucose before and after exercise (as soon as possible after exercise but no later than 15 minutes after the completion of exercise). For participant safety, capillary blood glucose must be 100 mg/dl or higher to begin exercise while the participant is at home. Participants will be given instructions to take additional carbs at the start

of exercise as needed based on CGM and trend to avoid post-exercise lows. See Appendix F outpatient instruction card. Participants will be required to have a person age 18 or older who attended a training session on the study present while the participant exercises at home. For at home exercise, the companion will stay after the exercise until the participant's CBG is > 100 mg/dl or for 60 minutes after exercise is completed, whichever is longer. Participants will inform study staff when exercising at home in the exMPC arm to allow for more direct remote monitoring. Staff will continue to monitor until exercise is completed and glucose is above 100 mg/dL.

Return to the Outpatient Clinic at OHSU

On the fourth day, participants will return to the AIMS lab in Biomedical Engineering Dept at CHH1, the CTRC clinic, or the Harold Schnitzer Diabetes Health Center in the morning. Participants will be asked to arrive early enough to provide sufficient time to allow the pods to be removed before the 72 hour mark of pod use. The study will be terminated and the participant's own insulin pump will be restarted. The study investigator will consult with the participant regarding appropriate insulin dosing for the remainder of the day. The Polar watch, Omnipod and Dexcom sensor will be removed from the participant. All infusion and sensor sites will be inspected for signs of irritation or infection. In addition, the sensor will be inspected for the possibility of breakage or fracture. If there is any evidence of sensor breakage, it will be recorded. If an area of inflammation of 1 cm or greater exists around the point of insertion, a de-identified photograph will be taken of the area and the participant will return 1-3 days later for a follow-up visit. A capillary blood glucose value will be taken immediately prior to discharging the participant. Participants will be given oral carbohydrate for values below 85 mg/dl, and will be given an insulin bolus if deemed appropriate by the study investigator for values above 150 mg/dl.

There is the option to complete this visit by Webex. Participants will be given shipping boxes for sending all devices back. Participants will connect with a study coordinator and investigator virtually to complete the visit.

If a study visit is stopped prematurely, such as due to technical problems, the participant will be asked if they can repeat the study visit that was terminated early with additional compensation provided. Repeating the study visit will be optional.

Hypoglycemia Treatment Guidelines

CBG < 70 mg/dl

- Give 15 grams of oral carbohydrate.
- Repeat treatment every 15 minutes as needed to raise blood glucose \geq 70 mg/dl.

Presence of STUPOR, LOSS OF CONSCIOUSNESS, or SEIZURE

- Give 1 mg glucagon IM
- Verify that insulin is turned off.
- Further management per study investigator.

Hyperglycemia Treatment Guidelines

If the sensed glucose is > 300 mg/dl, the participant will be instructed to check their blood glucose and to check the Omnipod for malfunction. This would include checking for insulin leaks, making sure Omnipod is securely adhered to skin, and check for error messages on the phone running the algorithm.

If capillary blood glucose value is over 300 mg/dl for more than 2 hours or is ≥ 400 mg/dl at any time, the participant will be instructed to check serum ketones using the Abbott Precision Xtra meter and to change out the Omnipod. If serum ketones are over 0.6 mM, the on-call study investigator will be alerted to discuss proper management, including delivering a correction bolus. In addition, the participant will be encouraged to drink sugar-free liquids. If serum ketones are above 1.5 mM at any time, the study will be stopped and insulin will be administered as directed by the on call investigator.

Cleaning and Disinfecting

All devices will be cleaned and disinfected between participants. The smart phone, Dexcom G6 transmitter, heart rate monitor and Omnipod PDMs are cleaned by study staff. Technicians who are disinfecting units will wash hands thoroughly and wear gloves. All items will undergo intermediate-level disinfection using Oxivir TB disposable wipes. The disinfectant will be applied and allowed to air dry. After disinfection, when the units are completely dry, they will be placed in a sealed bag labeled with participant information.

Stopping Rules

Individual study stopping rules

The closed-loop study will be stopped and open-loop control will be resumed under the guidance of the on call study investigator if any of the following occur after the start of the study: 1) capillary blood glucose falls to ≤ 40 mg/dl at any time point, 2) capillary blood glucose exceeds 400 mg/dl on two occasions (120 min or more apart within a 4 hour window), 3) capillary blood glucose exceeds 400 mg/dl on two occasions more than 120 minutes apart but outside of the 4 hour window and during that time, the capillary blood glucose has not fallen below 250 mg/dl, 4) serum ketones are above 1.5 mM at any time, 5) seizure or unconsciousness associated with hypoglycemia.

Entire study stopping rules

Triggers for reporting unanticipated problems are seizure, hospitalization, death or any other occurrence considered serious by the PI and Medical Monitor. If any studies are stopped for severe hypoglycemia or diabetic ketoacidosis, then the entire study will be halted. Severe hypoglycemia is defined as event that required the assistance of another person due to altered consciousness, and required another person to actively administer carbohydrate, glucagon, or other resuscitative actions. Diabetic ketoacidosis is classified as: symptoms such as polyuria, polydipsia, nausea, or vomiting, serum ketones >1.5 mmol/L or moderate/large urine ketones, either arterial blood pH<7.3 or venous pH<7.24 or serum bicarbonate <15, and treatment provided in a health care facility. If an event of severe hypoglycemia or diabetic ketoacidosis should occur, the study will be halted and a description of the serious adverse event and a new risk mitigation plan will be submitted to the FDA. The study will only resume once FDA approval is received. In addition, if there is any unexpected event such as death or patient hospitalization, the studies will be stopped until the root cause is evaluated.

The Fading Memory Proportional Derivative (exAPD) Algorithm

The exAPD algorithm determines insulin delivery rates based on proportional error, defined as the difference between the current glucose level and the target level, and the derivative error, defined as the rate of change of the glucose. Each of these errors is calculated over a time interval. The "fading memory" designation refers to weighting recent errors more heavily than remote errors. This weighting provides an adaptive component to the algorithm. In simple terms, the insulin rate is increased for high or rising glucose levels. Gain factors determine the degree to which proportional or derivative errors lead to changes in the delivery rate.

For the exAPD exercise-enabled mode, the exercise threshold will be set to 4 METs/min for every participant at the start of the study. If the corrected MET value is greater than 4 METs for a period of 5 consecutive minutes during the first exercise period, exercise is considered to be ongoing. An exercise dosing adjustment algorithm will be used when exercise is detected that has been previously tested and published. When exercise is detected while in the exAPD exercise-enabled mode, and the participant confirms he/she is exercising, insulin will be turned off for 30 minutes (nominal) immediately after detection of exercise. Subsequently, the insulin infusion rate will be reduced to 50% (nominal) for a period of 1 hour (nominal). The exercise detection algorithm will prompt the participant if exercise is occurring prior to adjusting dosing. For example, if the participant's METs exceed a threshold of 4 METs, the AP will detect this and ask the participant if they are exercising. If the participant acknowledges this and says "Yes", the AP will adjust the dosing. If the participant selects that they are not exercising, iPancreas will present a dropdown menu from which the participant can select their current activity for logging purposes (e. g. housework). It will also present several choices for the amount of time that exercise detection will be suspended so that the participant will not continuously receive detection alerts during that activity (15, 30 or 60 minute suspension). If the participant says "No", this is considered a false alert because the algorithm has detected exercise, but the participant was not actually exercising. Because these false alerts can be annoying to the participants, the AP includes an adjustable exercise detection threshold. The adjustable exercise detection threshold works as follows:

At the start of the study, the participants' exercise detection threshold will be set to 4 METs.

- On the first day of the study, when the participant exercises at the hospital, the AP records the participants' METs during exercise and also records the participants' METs during activities of daily living and other non-exercise events.
- Based on data from this controlled setting, a "lower bound MET" for that participant will be calculated based on a lower-bound confidence interval set around their METs recorded during exercise.
- If that participants' lower bound MET during exercise is greater than 4, their maximum allowable exercise threshold value (MAETV) will be set to the lower bound MET. Otherwise, the MAETV will remain at 4.
- Every time a false alert occurs for detecting exercise, the participants' exercise detection threshold will increase by 0.25 MET. However, the exercise detection threshold will never exceed the MAETV described above.

The Model Predictive Control (exMPC) Algorithm

Our exMPC algorithm uses a glucoregulatory model to predict glucose outcomes over a predicted horizon (Np), and mathematically solve for the optimal insulin doses across the control horizon (Nc) to bring the participant to target. The model is updated at each timestep by a Kalman filter, which uses the difference between CGM observations and model predictions to update the physiologic model states. In short, a model-predictive controller uses a physiologic model to calculate how much insulin is required to bring someone to glucose target, and these predictions are adapted to the specific participant using a Kalman filter.

For the exMPC exercise-enabled mode, there is a model within the controller that takes as an input the aerobic metabolic expenditure in addition to the CGM and meal inputs. The algorithm uses heart rate and accelerometer data collected on the patient's body to calculate metabolic expenditure. The metabolic expenditure then acts on the model for the insulin dynamics, whereby more energy expenditure and longer duration exercise can lead to a more substantial effect of insulin on the CGM (i.e. the CGM will drop more in response to more intense aerobic exercise and with longer duration exercise). In this way, the model within the exMPC control algorithm is always aware of exercise as a continuous input to the system and can respond dynamically to short or long, light, moderate, or intense exercise bouts. We expect that the exMPC controller will be able to dynamically adapt to either short or long exercise bouts. We also expect the exMPC to dynamically adapt to in-home exercise better, which can be more variable than static exercise that is performed within a clinic under controlled conditions.

Since the exMPC mode treats exercise differently than the exAPD mode, some exercise alerts to the user will not populate. The low glucose alert will populate when sensor glucose goes below 70 mg/dl, instead of 85 mg/dl in exAPD mode with exercise detection. The alert not allowing exercise when ketones are above 0.6 mM will not populate while in exMPC mode, but the participants will be coached not to exercise when ketones are high.

Statistical methods

The primary study endpoint is percent of time with glucose sensor <70 mg/dL during the inpatient clinic day. The hypothesis to be tested is the exMPC exercise-enabled closed-loop system will increase time in range as compared to the exAPD exercise-enabled closed-loop system. As our main approach, we will perform a two-sided Wilcoxon test for the paired differences. This test is widely used and easily interpreted. However, it cannot accommodate missing data or estimate crossover effects, and while we do not anticipate that these will affect our study outcomes, we plan to conduct a Supplementary analysis using a linear regression model with bootstrapped standard errors, which do not require distributional assumptions. In resampling for the bootstrap, we will use study subjects as the sampling unit to account for correlation between their repeated observations. The general model is as follows:

$$
(1) \tY_{ij} = \beta_0 + \beta_1 Trt_j + \beta_2 Seq_i + \beta_3 Period_j
$$

In this model, the outcome Y*ij* is the percent of time with glucose sensor <70 mg/dL for person *i* in observation period *j*. *Trtj* represents the treatment arm (exMPC=1, exAPD=0). *Seqi* is an indicator for sequence (AB vs BA) and *Periodj* indicates when *Trtj* occurred in the sequence. The coefficient *β⁰* represents the mean response for exAPD, and the coefficient *β¹* represents the difference between treatments to be tested. The coefficients *β²* and *β3*, mean differences for sequence or period, are both expected to equal zero; a large effect or significance test with p-value <0.05 will be considered evidence of a carryover effect. Hypothesis tests will be two-sided with type I error set to 0.05. Because statistical tests are specified and prioritized *a priori* and our proposed endpoints are highly related, we will follow recommendations to report p-values and confidence intervals rather than adjust for multiple comparisons.

Model fit and alternatives: Goodness of fit statistics and model fitting diagnostics will be used to assess for influential points and to evaluate alternative model specifications. If needed to compensate for some skew and heteroscedasticity in secondary outcomes, we will use bootstrapped variance estimators and compare these against robust estimates.

Missing data: Our primary approach will include all available data as an intention-to-treat analysis, regardless of whether study subjects completed both arms. An observation will be included if at least 24 hours of data are available. However, we expect very low levels of missingness; in our previous four-arm crossover study, only one subject withdrew before completing at least two arms. Dropped CGM values will be interpolated over short time periods, using capillary values for calibration when available. In the event of ≥10% missingness in either subject data or CGM values, we will analyze outcomes under multiple imputation to compensate for potential bias.

Additional considerations for selected secondary endpoints: The number of carbohydrate treatments will be modeled using model (1) above with Poisson regression for count data. The coefficient of variation of glucose is non-normal without an obvious probability distribution, so that bootstrapping will be our primary approach for this endpoint. Data for percent of time with sensed glucose <54 mg/dl may be sparse; if so, we will present counts of events where glucose went below this level, counting as separate events when the time between them is ≥ 30 minutes.

Confidentiality and Protection of Human Participants

RISKS and BENEFITS

Risks: The risks of the protocol procedures are considered minor. Nonetheless, since pumps and sensors used within automated glucose control systems are imperfect, there is a risk for hyperglycemia and hypoglycemia. All studies will issue alerts and will be remote monitored during each visit with unserviced alerts being pushed to the study coordinator and investigator. A study investigator will be on call at all times.

Risks from exercise include falls, sprains, bruises, very low risk of bone fractures and head trauma. The likelihood of significant harm is quite low.

Rarely, there can be allergic responses to insulin, such as skin redness, hives, itching of the skin, swelling of the mouth, or breathing difficulties. These reactions are considered very unlikely.

The following events have been identified as possible anticipated device-related adverse events of Dexcom G6 sensor insertion and wear:

- Excessive pain or discomfort from either system deployment or during wear period (8 or greater on a 10-point Likert scale)
- Excessive bleeding, defined as requires removal of the device to stop bleeding
- Hematoma, defined as induration at the sensor insertion location (ecchymosis is a known consequence of needle skin puncture or pressure from sensor pod and will not be captured as an AE)
- Edema from sensor and/or adhesive tape that is significant and non-resolving within 48 hours of sensor pod removal
- Erythema from sensor and/or adhesive tape that is significant and non-resolving within 48 hours of sensor pod removal
- Local infection, defined as presence of pus at either sensor wire or sensor pod site
- Sensor or introducer needle fracture during insertion/wear/removal For this reason, the study investigator will inspect each removed sensor for the possibility of breakage or fracture. Any evidence of sensor breakage will be recorded and reported to FDA and the sensor company*.*

Benefits: The participant may not directly benefit from being in this study; however, their participation may help to advance automated insulin delivery technology.

COSTS:

The participants who stay in clinic each day for the exMPC arm will receive \$1025. If a participant withdraws early from the study, compensation will be given as follows: \$50 for run-in, \$225 per in clinic day for the exMPC arm, \$300 for the exAPD arm.

Otherwise, participants will receive \$650 for completion of all study visits. If a participant withdraws early from the study, compensation will be given as follows: \$50 for run-in, \$300 for the exMPC arm and \$300 for the exAPD arm. There is no compensation for the screening visit. If a participant is asked to repeat a study due to technical problems, he/she will receive an additional \$300.

Monitoring Entity:

This investigation will be monitored by the co- investigator Leah Wilson MD. Dr. Wilson has no commercial interest in any of the companies which manufacture any of the devices used in this study. Drs. Jessica Castle, Peter Jacobs and Joseph El Youssef are inventors on patents regarding the algorithms.

Data Collection:

Participant privacy will be protected by using a three-digit identifying number to code study documents. All paper source documents will be kept in a locked cabinet for a minimum of five years. Data for this project will be stored in an AWS server developed by our group that has undergone a security review.

Recording of Data:

Investigators and staff will record data collected during the clinical trial on the CRF's. The CRFs will include:

- 1. Screening form
- 2. Dexcom G6 Training Visit
- 3. iPancreas Training
- 4. Companion Training
- 5. Day 1 Inpatient exMPC Closed-loop Study Visit
- 6. Day 2 Inpatient exMPC Closed-loop Study Visit
- 7. Day 3 Inpatient exMPC Closed-loop Study Visit
- 8. Day 1 Inpatient exAPD Closed-loop Study Visit
- 9. Day 4 exMPC Closed-loop Study Visit
- 10. Day 4 exAPD Closed-loop Study Visit
- 11. Phone Update Form
- 12. Adverse Event form
- 13. Serious Adverse Event form
- 14. Concomitant Medications

The Principal Investigators may authorize other personnel to make entries in the CRF. The coded data collected during this study will be used for analysis of the primary and secondary endpoints listed in this protocol. The key to the code for this study will not be stored in the repository and only named study members on this project will have access to the key for this study. Researchers who request data from the repository will not receive any identifiers aside from date and we do not anticipate that the date will allow those researchers to re-identify the data. However, some of the researchers named on this project may use the data from the repository which would mean that the repository data will still be potentially identifiable to those who have access to the key as part of this project. The coded data will also be stored in the OregonAPC repository according to IRB protocol 19858. During screening, all new participants will sign the consent form to store their study data in the data repository. The data to be collected includes: 1) glucose sensor data, 2) blood glucose data, 3) insulin data, 4) physical activity data, and 5) food and exercise data. All data, except for blood glucose, is aggregated by the iPancreas app. The blood glucose data is collected through downloading the Contour Next BG meters and exporting data as an excel file. There are no biological specimens collected during this study.

Monitoring Procedures:

This protocol is written in accordance with the principles established by the 18th World Medical Assembly General Assembly (Helsinki, 1964) and amendments and clarifications adopted by the 29th (Tokyo, 1975), 35th (Venice, 1983), 41st (Hong Kong, 1989), 48th (Somerset West, South Africa, 1996), 52nd (Edinburgh, 2000), 53rd (Washington, 2002),

55th (Tokyo, 2004), 59th (Seoul, 2008), and 64th (Brazil, 2013) General Assemblies. The investigator will ensure that the study described in this protocol is conducted in full conformance with those principles, the protocol, current FDA regulations, ICH Good Clinical Practices (GCP) guidelines, Good Laboratory Practices (GLP) guidelines, local ethical and regulatory requirements, including the Federal Food, Drug and Cosmetic Act, U.S. applicable Code of Federal Regulations (title 21), any IEC requirements relative to clinical studies.

Should a conflict arise, the investigator will follow whichever law or guideline affords the greater protection to the individual participant. The investigator will also ensure thorough familiarity with the appropriate use and potential risks of use of the study device, as described in this protocol, prior to the initiation of the study.

Adverse Event Reporting

At all study visits, study staff will determine if any adverse events (AEs) have occurred. Disease related events that are chronic in nature and occur as part of the progression of the diabetes disease state (i.e. diagnoses of retinopathy, nephropathy, and neuropathy) will not be captured as adverse events in this study. Hypo- and hyperglycemia will not be considered AEs unless participant has positive ketones or displays symptoms of hypoglycemia such as: loss of consciousness, slurred speech, hospitalization or EMS services called.

One of the investigators will write up a description of the adverse event/unanticipated problem. All reportable new information (RNI) will be reported to the IRB within five calendar days after the PI learns of the event. RNI is any information that might meet the regulatory definition of an unanticipated problem involving risks to subjects or others or serious or continuing noncompliance that might impact the criteria for IRB approval. The report will be submitted to the IRB by the principal investigator or study coordinator. A summary of all UP's and adverse events, including those that do not meet the requirement for RNI, will be submitted with the continuing review. All AEs will be monitored until adequately resolved or stable. Information regarding AEs that occur during the study will be entered into appropriate CRFs. Such information will include, at a minimum:

- Date of event
- Severity
- Outcome
- Resolution of event

Definition of Adverse Event

An AE is any clinically significant undesirable experience (sign, symptom, illness, or other medical event) meeting the causality definition above that appears or worsens in a participant during a clinical study. A clinically significant event is any event (sign, symptom, lab/imaging abnormality, or diagnosis) that is noteworthy enough to merit documentation in standard medical records (e.g. history and physical, progress notes, clinic visit notes, etc.). Other non-clinically significant events (e.g. colds, minor headaches, etc.) *may* be documented on the comments CRF. Mild hypoglycemia is expected in persons with diabetes using insulin and are typically self-limiting in nature; thus, this will not be captured as an AE. The Medical Monitor will have the final say in determining the causality. Adverse events that continue after the participant's discontinuation or completion of the study will be followed until their medical outcome is determined or until no further change in the condition is expected.

Serious Adverse Events

A serious adverse event (SAE) is defined as any untoward medical occurrence that:

- Results in death
- Is life-threatening; (substantial risk of dying at the time of the adverse event or suspicion that continued use of the device would result in a participant's death
- Requires inpatient hospitalization or prolongation of existing hospitalization
- Results in persistent or significant disability/incapacity or substantial disruption of the ability to conduct normal life functions
- Requires medical or surgical intervention to prevent permanent impairment or damage.

Exceptions to the SAE definition will include the following:

- Elective surgery
- A planned hospitalization for pre-existing condition, without a serious deterioration in health

Any SAE, including death, due to any cause (related or unrelated to the device), that may occur during a clinical study will be reported to the PI and Medical Monitor immediately (within 1 working day of learning of the event).

Severity of Adverse Events

The following definitions may be used to rate severity of AEs:

• **Mild**

Awareness of signs or symptoms, but easily tolerated; are of minor irritant type that is outside the norm for the disease state or subject; causing no loss of time from normal activities; symptoms would not require medication or a medical evaluation; signs and symptoms are transient. Mild hypoglycemia is expected in diabetes and will not be captured as an AE.

Example: ketosis not requiring an ER visit.

• **Moderate**

Discomfort severe enough to cause interference with usual activities, requiring treatment due to cognitive impairment, by family member or emergency personnel

Example: hypoglycemia with inability to self-treat, requiring third party assistance for treatment and/or an emergency room visit.

• **Severe**

Incapacitating, causing inability to do work or usual activities; signs and symptoms may be of systemic nature or require medical evaluation and/or treatment at a health care facility *Example: hypoglycemia with loss of consciousness or seizure involvement*

Relationship of Adverse Events to Study or Study Device

The investigator will categorize the relationship of the event to the study or study device as follows:

• **Not Related**

AE is due to an underlying disease state or concomitant medication or therapy not related to the device, disease or study.

• **Unlikely Related**

AE has minimum or no temporal relationship to the study device, disease or study participation and/or a more likely alternative etiology exists.

• **Possibly Related**

AE has a strong temporal relationship to the study device, disease or study procedures and alternative etiology is equally or less likely compared to the potential relationship to the device, disease or study.

• **Probably Related**

AE has a strong temporal relationship to the study device, disease or study and another etiology is unlikely.

• **Definitely Related**

AE has a strong temporal relationship to the study device, study procedures or disease and another etiology does not exist.

Unanticipated Problems

Unanticipated problems, including study, disease or device-related problems will be detected by reviewing descriptions of known or foreseeable adverse events and risks in the IRB-approved research protocol and the current IRB approved consent form, any underlying disease or conditions of the subject experiencing the adverse event, a careful assessment of

whether the adverse event is related or possibly related to the subject's participation in the study or if root cause or associations is with study devices.

Unanticipated Adverse Device Effect (UADE)

An unanticipated adverse device effect (UADE) is not expected to occur. An UADE is defined as any serious adverse effect on health or safety or any life-threatening problem or death caused by – or associated with – the device, if that effect, problem, or death was not previously identified in nature, severity, or degree of incidence in the investigational plan (including documents such as the protocol, the informed consent document, other study-related documents), or any other unanticipated serious problem associated with the device that relates to the rights, safety, or welfare of participants.

During the review of a reported SAE, if the PI and Medical Monitor input determines the severity or extent of the event was not cited in this protocol or associated protocol materials, and the event was classified as, 'possibly related' to the device, the event will be documented as an UADE. If the event is classified as an UADE, the Investigator must notify the IRB and Dexcom will notify the FDA within ten (10) working days of the original SAE notification.

If determined that the UADE presents an unreasonable risk to participants, we will terminate all investigations or parts of investigations presenting that risk as soon as possible, but not later than 5 working days after such determination is made and not later than 15 working days after we first receives notice of the original SAE. We will not resume a terminated study without IRB and FDA approval.

MDR Reportable Events/MDR Reporting

A device issue, whether related to a complaint or not, is an allegation from the participant or study personnel regarding an indication of the failure of a device to meet user expectations for quality or performance specifications. Device issues will be recorded onto appropriate CRFs by site personnel. The CGM and Omnipod devices are currently marketed. Therefore, the PI will follow the required reporting regulations to Dexcom or Insulet if an MDR reportable event occurs.

MDR reportable events are events that manufacturers become aware of that reasonably suggest one of their marketed devices may have caused or contributed to a death or serious injury, or has malfunctioned and the malfunction of the device would likely cause or contribute to a death or serious injury if the malfunction were to recur (21 CFR 803.3).

Confidentiality Procedures:

To protect confidentiality, standard institutional practices will be followed as described by the OHSU Information Security Directives at the following link: http://www.ohsu.edu/xd/about/services/integrity/policies/ipspolicies-info-secdirectiv.cfm#results to maintain the confidentiality and security of data collected in this study. Study staff will be trained with regard to these policies.

Paper files will be stored in locked filing cabinets in restricted access offices at site. After the study, source documents will be maintained at the participating clinical center (or offsite record storage facilities) 2 years after a marketing application is approved for our group's decision support device or discontinuance of pursuit of marketing approval. Electronic data will be stored on encrypted: computers, laptops and study smartphones. Electronic data is stored in OneDrive. Sites will be invited to upload data using a password protected folder in Microsoft OneDrive. iPancreas data will be housed in a custom cloud database on an OHSU secure server called the GRM (see below). Access to data/specimens is restricted to study personnel. Access to data requires OHSU ID/password authentication See IRB protocol 19858 for a complete description of the confidentiality and security of the study data collected during this study to be stored in the OregonAPC repository.

iPancreas Guidance Remote Monitoring (GRM) Cloud Server

All of the data collected will be streamed over the Internet (using secure sockets encryption) to an OHSU secure instance of an AWS cloud storage server every 5 minutes. Authentication between the phone and the AWS server is done using OAuth2. Data transmitted between the phone and the AWS server is encrypted using HTTPS/SSL. The code managing authentication and data transfer is Python version 3.7.0. Data acquired from the app is displayed via a physician web portal. The physician web portal user interface is written in Javascript version 1.8.5. There is no personally identifiable data stored with the data sent to the AWS server. The server shall be capable of receiving the following types of data (1) CGM data, (2) blood glucose and ketone data, (3) insulin dosing data, (4) insulin on board, (5) alerts, (6) exercise data,

and (7) settings. All types of data shall be indexed by participant ID and by date/time. Data shall be stored on the server in a secure database. Each data packet shall be accompanied by an authentication identifier determined through oauth. The AWS server has undergone a security review by OHSU IPS.

Appendix A: Physical Activity Readiness Questionnaire

Physical Activity Readiness Questionnaire (PAR-Q) and You

Regular physical activity is fun and healthy, and increasingly more people are starting to become more active every day. Being more active is very safe for most people. However, some people should check with their doctor before they start becoming much more physically active.

If you are planning to become much more physically active than you are now, start by answering the seven questions in the box below. If you are between the ages of 15 and 69, the PAR-Q will tell you if you should check with your doctor before you start. If you are over 69 years of age, and you are not used to being very active, check with your doctor.

Common sense is your best guide when you answer these questions. Please read the questions carefully and answer each one honestly:

Informed use of the PAR-Q: Reprinted from ACSM's Health/Fitness Facility Standards and Guidelines, 1997 by American College of Sports Medicine

Appendix B: Devices

Insulet Omnipod insulin management system which includes PDM and Omnipod

Dexcom G6 Continuous Glucose Monitoring System which includes Sensor and Sensor Transmitter

Contour Next Blood Glucose Meter Abbott Precision Xtra Meter

Appendix C: Hypoglycemia Awareness questionnaire: This survey item will be used to categorize awareness or having reduced awareness of hypoglycemia.

1. Check the category that best describes you: (check one only)

- \Box I always have symptoms when my blood sugar is low (A)
- \Box I sometimes have symptoms when my blood sugar is low (R)
- \Box I no longer have symptoms when my blood sugar is low (R)

2. Have you lost some of the symptoms that used to occur when your blood sugar was low?

- \Box Yes (R)
- \Box No (A)

3. In the past 6 months how often have you had moderate hypoglycemia episodes? (Episodes where you might feel confused, disoriented, or lethargic and were unable to treat yourself).

- \Box Never (A)
- \Box Once or twice (R)
- \Box Every other month (R)
- \Box Once a month (R)
- \Box More than once a month (R)

4. In the past year, how often have you had severe hypoglycemia episodes? (Episodes where you were unconscious or had a seizure and needed glucagon or intravenous glucose?)

- \Box Never (A)
- \Box 1 time (R)
- \Box 2 times (R)
- \Box 3 times (R)
- \Box 4 times (R)
- \Box 5 times (R)
- \Box 6 times (R)
- \Box 7 times (R)
- \Box 8 times (R)
- \Box 9 times (R)
- \Box 10 times (R)
- \Box 11 times (R)
- \Box 12 or more times (R)
- 5. How often in the last month have you had readings < 70 mg/dl with symptoms?
	- □ Never
	- \Box 1 to 3 times
	- \Box 1 time/week
	- \Box 2 to 3 times/week
	- \Box 4 to 5 times/week
	- Almost daily
- 6. How often in the last month have you had readings < 70 mgdl, without symptoms? R: 5<6, A: 6<5;
	- Never
	- \Box 1 to 3 times
	- \Box 1 time/week
	- \Box 2 to 3 times/week
	- \Box 4 to 5 times/week
	- Almost daily
- 7. How low does your blood sugar need to go before you feel symptoms?
	- \Box 60-69 mg/dl (A)
	- \Box 50-59 mg/dl (A)
	- \Box 40-49 mg/dl (R)
	- \Box < 40 mg/dl (R)
- 8. To what extent can you tell by your symptoms that your blood sugar is low?
	- \Box Never (R)
	- \Box Rarely (R)
	- \Box Sometimes (R)
	- \Box Often (A)
	- \Box Always (A)

Appendix D: Alert Manager Specifications

Appendix E: Exercise Video Outline

Appendix F: **Instruction cards for participants to take additional carbs at the start of exercise as needed based on CGM and trend to avoid post-exercise lows**

Inpatient instruction card:

Outpatient instruction card:

IPANCREAS GUI

6 IN-HOME AND IN-CLINIC EXERCISE CARBOHYDRATE INTAKE AND GLUCOSE OUTCOMES.

6.1 IN-HOME CARB AND CGM EXERCISE RESULTS

Supplementary Table 6.1 exMPC In-home exercise carbohydrate intake and CGM (note that not all participants reported their in-home exercise).

Supplementary Table 6.2: exAPD in-home exercise carbohydrate intake and CGM (note that not all participants reported their in-home exercise).

6.2 IN-CLINIC CARB AND CGM EXERCISE RESULTS

Supplementary Table 6.3 exMPC in-clinic exercise carbohydrate intake and CGM

Supplementary Table 6.4: exAPD in-clinic exercise carbohydrate intake and CGM

7 INSULIN ON BOARD AT THE START OF EXERCISE

We explored whether automated insulin delivered in the two hours prior to exercise was different for exAPD compared with exMPC. We found that the automated insulin on board at the start of exercise was 1.89 U [0.4 2.4] for exAPD and 1.4 U [0.2 2.4] for exMPC with no statistically significant difference as determined by a Wilcoxon signed rank test (P=.7). Supplementary Figure 7.1 below shows the distribution of automated insulin on board delivered in the two hours leading up to in-clinic exercise.

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