

Personalised adaptive basal-bolus algorithm using SMBG/CGM data



A Mobile Platform for Personalization of Insulin Delivery based on a Patch Pump and Reinforcement Learning

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Background and Aims

An essential component of the diabetes management plan is the glucose monitoring using either meters for self-monitoring of blood glucose (SMBG) or more recently continuous glucose monitors (CGM).

The main objective of the study is the **simultaneous personalisation of the basal and bolus insulin** to be delivered by a pump **independent of the used glucose monitoring device (SMBG or CGM)** using a newly proposed **adaptive basal-bolus algorithm (ABBA)**.

Method

ABBA is based on reinforcement learning (RL), a branch of artificial intelligence algorithms, that

"...allows the system to learn its behaviour based on feedback from the environment"

After an initialization phase of seven days using patient's CGM and insulin pump data the ABBA estimates the basal and bolus independently if he/she is using CGM or SMBG.

ABBA

Input: Glucose concentrations (from SMBG or CGM) of day i

Output: Basal rate for day $i+1$, and three Carbohydrate to Insulin Ratios (CIRs) for

- breakfast - CIR_B,
- lunch - CIR_L, and
- dinner - CIR_D

for the day $i+1$

In silico evaluation and daily scenario

The training version of the FDA accepted UVA/Padova T1DM Simulator v3.2

- The scenario lasted for 98 days (Days 2-8 for initialization; Days 9-98 for control).
- Three meals (breakfast, lunch and dinner) were considered with variabilities in the announcement of mealtime (± 15 min) and carbohydrate amount (± 10 g for main meals and ± 5 g for snack). In addition, an uncertainty in the order of $\pm 50\%$ was introduced to simulate the error of patient's carbohydrate content self-estimations.
- Variable insulin sensitivity (dawn phenomenon -50% from 4:00 AM to 8:00 AM and interday variability in the order of $\pm 25\%$).

Both variabilities and uncertainties follow uniform distributions.

Results

	BG (mg/dL)	% in target range	% in Hypo	% in Hyper	Total daily insulin (U)
<i>(mean \pm standard deviation)</i>					
Adults					
ABBA _{CGM}	140.9 \pm 18.4	87.5 \pm 16.1	1.0 \pm 1.1	11.5 \pm 15.4	43.2 \pm 10.8
ABBA _{SMBG}	143.5 \pm 18.9	86.9 \pm 16.7	0.6 \pm 0.9	12.5 \pm 16.0	42.6 \pm 9.9
Adolescents					
ABBA _{CGM}	148.1 \pm 11.5	75.7 \pm 12.2	2.4 \pm 2.0	21.9 \pm 12.3	31.6 \pm 6.7
ABBA _{SMBG}	145.8 \pm 9.3	77.8 \pm 13.6	2.6 \pm 2.2	19.6 \pm 12.7	31.8 \pm 7.1
Children					
ABBA _{CGM}	149.3 \pm 9.0	75.0 \pm 9.8	1.8 \pm 1.6	23.2 \pm 8.8	15.9 \pm 3.8
ABBA _{SMBG}	150.5 \pm 10.8	75.2 \pm 12.4	1.1 \pm 1.3	23.7 \pm 11.5	15.9 \pm 3.7

No statistically significant difference was observed between ABBA_{CGM} and ABBA_{SMBG}, $\alpha=0.05$
Target range: [70 180] mg/dl; Hypo < 70 mg/dl; Hyper > 180 mg/dl

Conclusions

ABBA, the proposed RL algorithm, demonstrated its ability to achieve

- glucose control in an *in silico* population using a complex scenario characterized by a high degree of variabilities and uncertainties,
- comparable performances for both CGM and SMBG, without affecting the total daily insulin dose.

Next step

ABBA within the framework of a clinical study