



Smart Technologies for Biomedical and Environmental Applications A joint DEWS@UNIVAQ – IASI@CNR Laboratory



Model-based control of plasma glycemia: in quest of robustness

Pasquale Palumbo^{1,3}, Mario Di Ferdinando², Alessandro Borri³

A. De Gaetano³, M.D. Di Benedetto², S. Panunzi³, P. Pepe², G. Pola²,

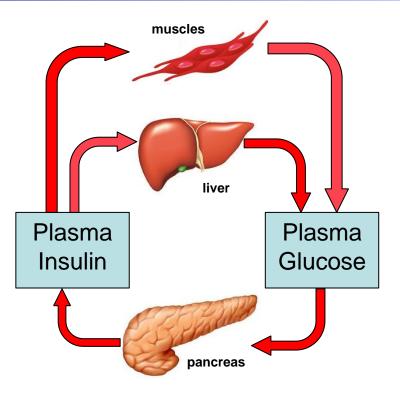
Department of Biotechnologies and Biosciences, University of Milano-Bicocca, Milan, Italy
 Department of Information Engineering, Computer Science and Mathematics,

University of L'Aquila, L'Aquila, Italy

3 National Research Council of Italy (CNR-IASI), Rome, Italy

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Physiological Glucose Control



High levels of glucose concentration (e.g. after a meal) stimulate **pancreatic insulin release** that:

- enhance glucose uptake in muscles
- allows the liver to storage extra glucose (as glycogen)

Glucose is the main energy source for the cells

Its basal concentration needs to be constrained within a narrow interval [60-90]mg/dl

Plasma glucose concentration is kept under control (mainly) by means of insulin hormone

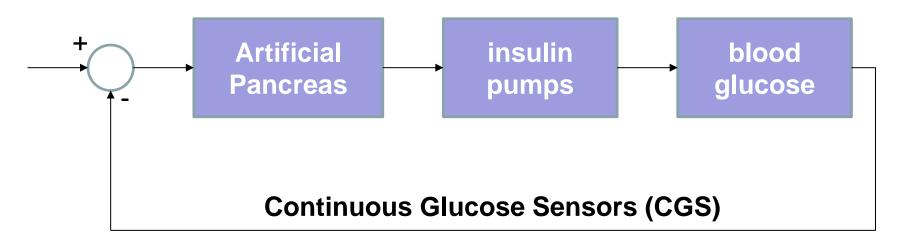
Diabetes comprises metabolic disorders characterized by hyperglycemia resulting from impaired insulin secretion and/or action

- Type 1 Diabetes Mellitus (T1DM): absolute deficiency of insulin secretion
- Type 2 Diabetes Mellitus (T2DM): resistance to insulin action and/or inadequate insulin secretory response

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Control Theory meets Glucose Control

Artificial Pancreas: refers to the set of glucose control strategies required for diabetic people and delivered by means of exogenous insulin administration



AP task: to close the loop automatically, safely, without any patient operation

Subcutaneous injections:	Intravenous infusions:
- more widespread, since the dose is administered by the patients	 rapid delivery with negligible delays
themselves	- more technology and a direct
 modeling the absorption from the subcutaneous depot 	supervision of a physician (usually adopted in ICU)

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Model-less vs Model-based approach

Model-less approach: no models of the glucose-insulin system are considered to design the control law

- It provides a control rule for the insulin infusion rate mainly based on experimental data
- The controllers mimic the pancreatic glucose response
- Its successful working often relies on human expertise in the adjustments of the algorithm parameters to different patients

Model-based approach: the control law is designed by exploiting the model of the glucose-insulin system.

- The control problem may be treated mathematically and optimal strategies may be determined
- It allows to test *in-silico* a control strategy before involving patients
- It requires a preliminary identification step to adapt the model to a given patient

The AP: State of the art

> AP for T1DM:

- many model-less approaches (e.g. PID, Fuzzy Logic, Model Predictive Control), most validated in closed-loop on a T1DM comprehensive model (UVA/Padua simulator, accepted by the FDA as a substitute of animal trials)
 L. Magni, G. De Nicolao (Pavia), B. Kovatchev (Virginia), J. Doyle III (California)
- model-based approaches, usually exploiting Model Predictive Control
 - R. Hovorka (UK), Moog (CNRS)

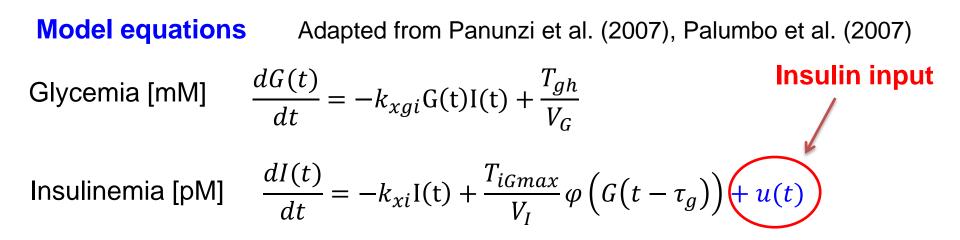
> OUR contribute, AP for T2DM:

- model-based approach: we exploit a Delay Differential Equation (DDE) system to model the endogenous insulin delivery rate
- we exploit glucose measurements to infer real-time estimates of the plasma insulin concentration (possibly by means of observers)
- the control law is designed by means of glucose measurements (and insulin estimates)
- the control law is validated by closing the loop on a modified version of the UVA/Padua simulator

Challeges in the AP

- From a control-theoretic viewpoint, insulin is a non-negative input (it reduces glycemia)
- ✓ Dealing with secretion delays
- ✓ Food as a source of uncertainty
- ✓ Random variations (hormones, stress, physical activity...)
- The subcutaneous compartment introduces filtering/delay effects (Insulin On-Board, IOB)
- ✓ Continuous-discrete system

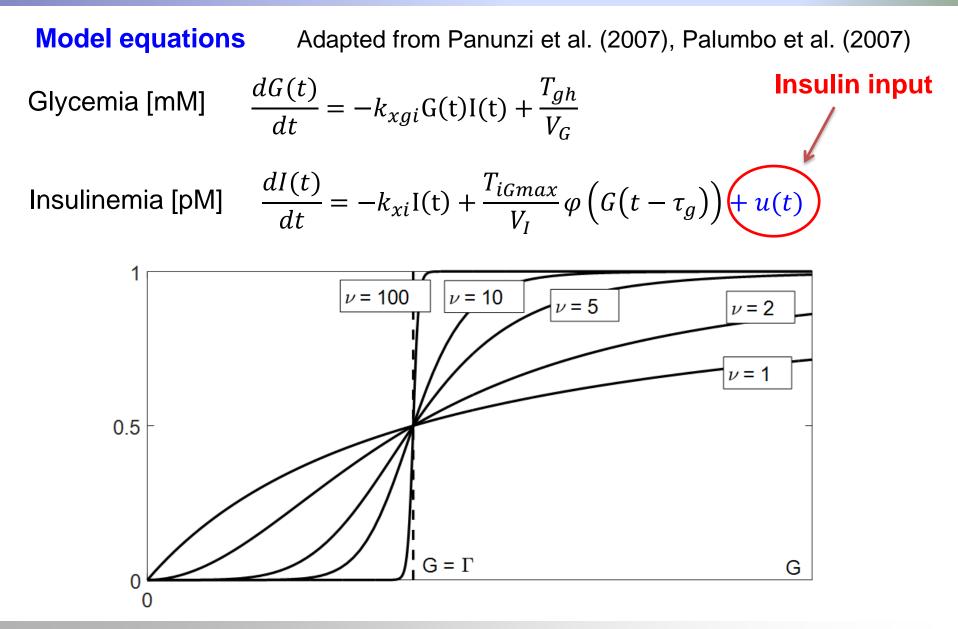
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Parameters • k_{xgi} : insulin sensitivity index

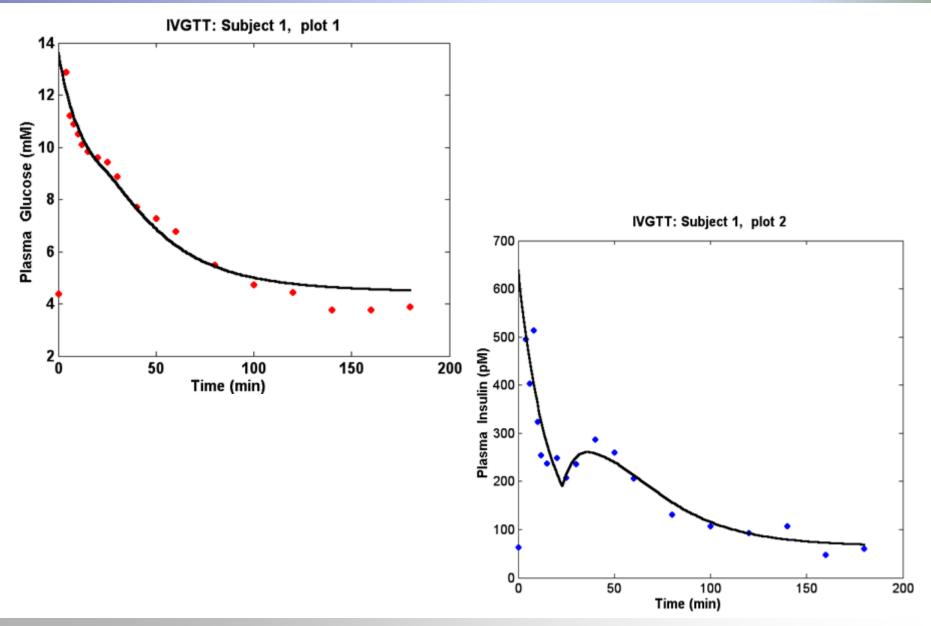
- T_{gh} : net balance between hepatic glucose output and insulin-independent zero-order glucose tissue uptake
- V_G , V_I : apparent distribution volume for glucose, insulin
- k_{xi} : first-order disappearance rate constant for insulin
- τ_g : apparent delay with which the pancreas varies secondary insulin
- release in response to varying plasma glucose concentrations
- T_{iGmax} : maximal rate of second-phase insulin release

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- It is mathematically well-posed
 - Single locally attractive equilibrium (basal conditions)
 - Positive, limited solutions
 - Global stability guaranteed under conditions on parameters
 - Physiologically limited pancreatic secretion ability
- It is statistically robust
 - Parameters statistically identifiable with very good precision via standard perturbation experiments, such as IVGTT
- It is a Compact Model
 - "minimal" set of independent (physiological) parameters



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Model-based control

Nonlinear feedback control for time-delay systems

The aim of control is to **reduce a high basal glycemia** to a lower level, according to a **reference glucose trajectory**, by means of intravenous/subcutaneous insulin administration

We **do not** want:

- dangerous glucose oscillations: avoid hypoglycemias
- not physically implementable control laws (avoid negative insulin infusions)

Therefore:

- the reference glucose trajectory needs to be slow enough
- in-silico simulations are essential to synthesize/validate an effective control law

Need for **robustness** (relying on fitting the model on a patient)

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Time-Delay Model-Based Control of the Glucose–Insulin System, by Means of a State Observer

Pasquale Palumbo^{1,*}, Pierdomenico Pepe², Simona Panunzi¹, Andrea De Gaetano¹

¹ Istituto di Analisi dei Sistemi ed Informatica "A. Ruberti", Consiglio Nazionale delle Ricerche (IASI-CNR), BioMatLab - UCSC - Largo A. Gemelli 8, 00168 Rome, Italy;
² Dipartimento di Ingegneria Elettrica e dell'Informazione, Università degli Studi dell'Aquila, 67040 Poggio di Roio, L'Aquila, Italy

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Model-based control of plasma glycemia: Tests on populations of virtual patients



P. Palumbo^{a,*,1}, G. Pizzichelli^{b,c,1}, S. Panunzi^a, P. Pepe^d, A. De Gaetano^a

* Istituto di Analisi dei Sistemi ed Informatica "A. Ruberti", Consiglio Nazionale delle Ricerche (IASI-CNR), Bio MatLab - UCSC - Largo A. Gemelli 8, 00168 Roma, Italy

^bIstituto Italiano di Tecnologia, Center for Micro-BioRobotics@SSSA, Viale R. Piaggio 34, 56025 Pontedera, Italy

⁶ Scuola Superiore Sant'Anna, The BioRobotics Institute, Viale R. Piaggio 34, 56025 Pontedera, Italy

^dUniversità degli Studi dell'Aquila, 67040 Poggio di Roio, L'Aquila, Italy

Luenberger-Like **Observers** for Nonlinear **Time-Delay Systems** with Application to the Artificial Pancreas

THE ATTAINMENT OF GOOD PERFORMANCE

ALESSANDRO BORRI, FILIPPO CACACE, ANDREA DE GAETANO, ALFREDO GERMANI, COSTANZO MANES, PASQUALE PALUMBO, SIMONA PANUNZI, and PIERDOMENICO PEPE

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Semiglobal Sampled-Data Dynamic Output Feedback Controller for the Glucose–Insulin System

Mario Di Ferdinando[®], Pierdomenico Pepe[®], Pasquale Palumbo, Simona Panunzi[®], and Andrea De Gaetano[®]

2019 18th European Control Conference (ECC) Napoli, Italy, June 25-28, 2019

Symbolic models approximating possibly unstable time-delay systems with application to the artificial pancreas

Giordano Pola, Alessandro Borri, Pierdomenico Pepe, Pasquale Palumbo and Maria D. Di Benedetto

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