

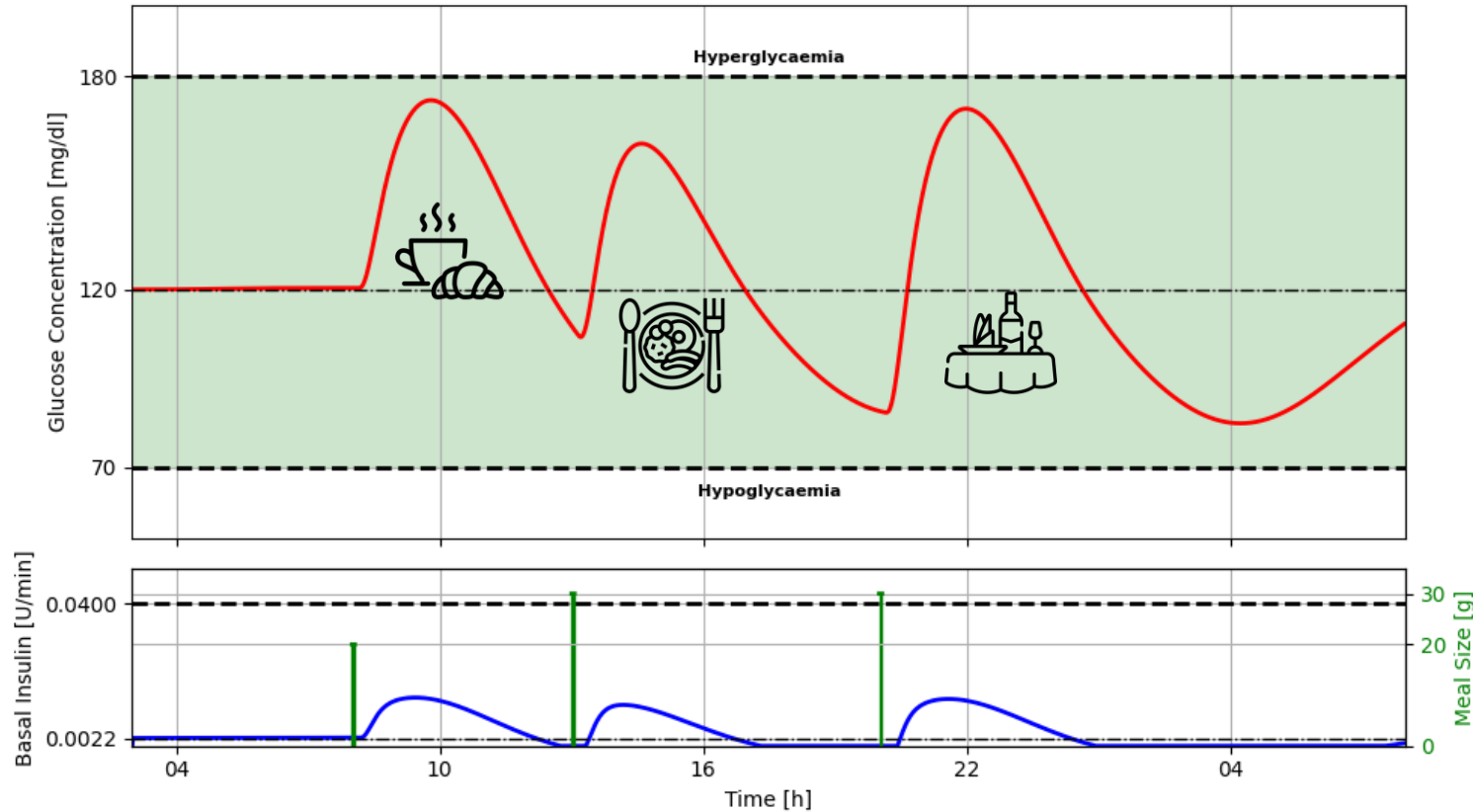


The illustration shows a person's midsection. On the left, an insulin pump is attached to the skin, with a tube leading to a cannula. On the right, a CGM sensor is attached to the skin, with red concentric arcs indicating signal transmission. A blue belt is visible at the bottom.

# Artificial Pancreas Controllers

Giacomo Mastroddi - InsulinCo

## PID Controller - Current State



### Meal intakes

The current scenario involves a breakfast with 20 grams of carbohydrates, and lunch and dinner with 30 grams each.

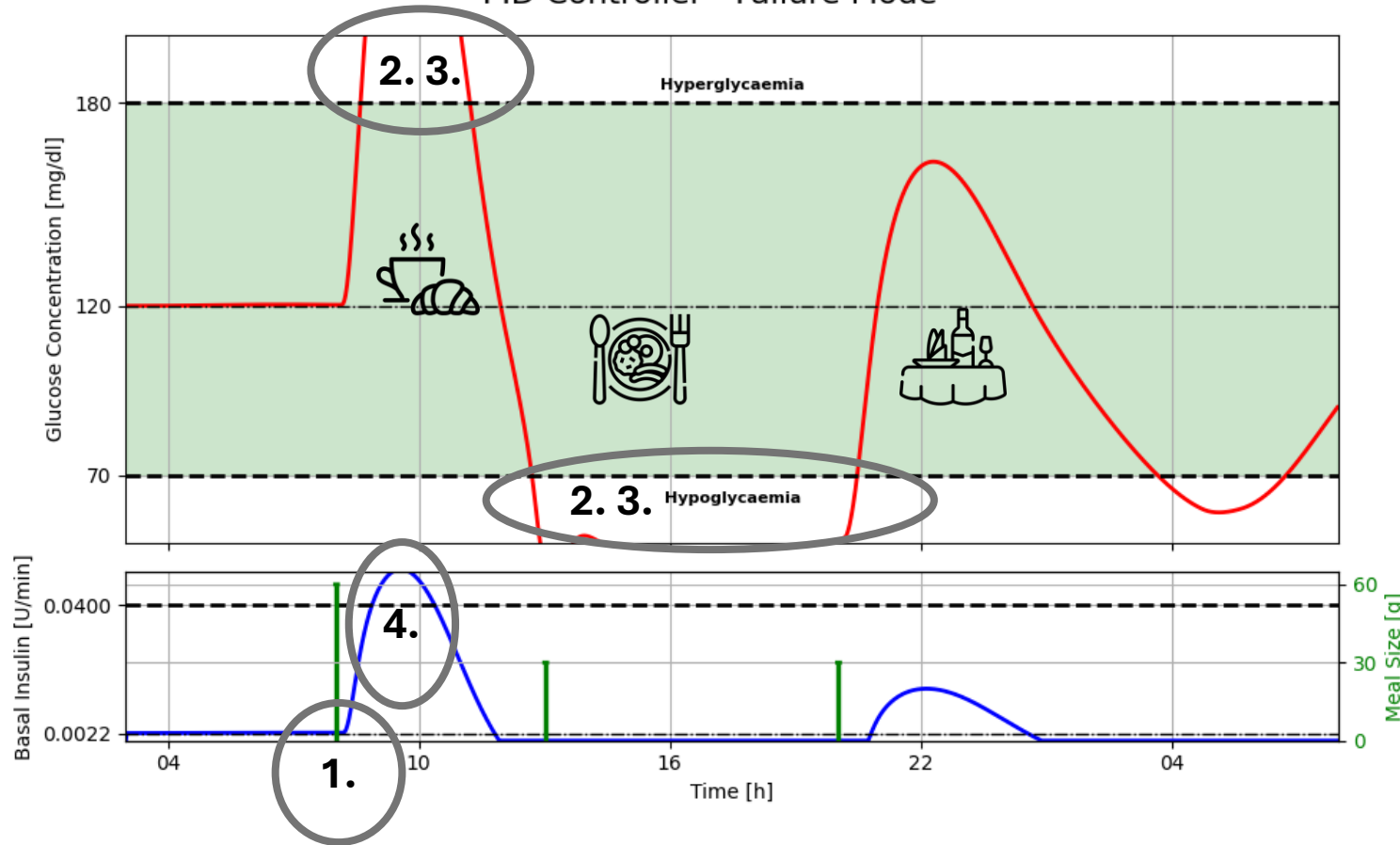
### Controller

A PID controller is used, which responds to the error signal between the measured current state and the reference state.

### Behaviour

- Insulin injections are administered only after detecting a deviation from the reference glucose level.
- With these small meal intakes, administering insulin after the glucose level rises keeps the patient's glucose within the safe range.
- The response is moderate, preventing the insulin pump from reaching its upper limit of 0.04 U/min.

## PID Controller - Failure Mode



## Conclusion

- Prolonged violation of glucose limits can lead to hospitalization or even death of the patient.
- A controller with predictive capabilities and the ability to satisfy constraints is essential.



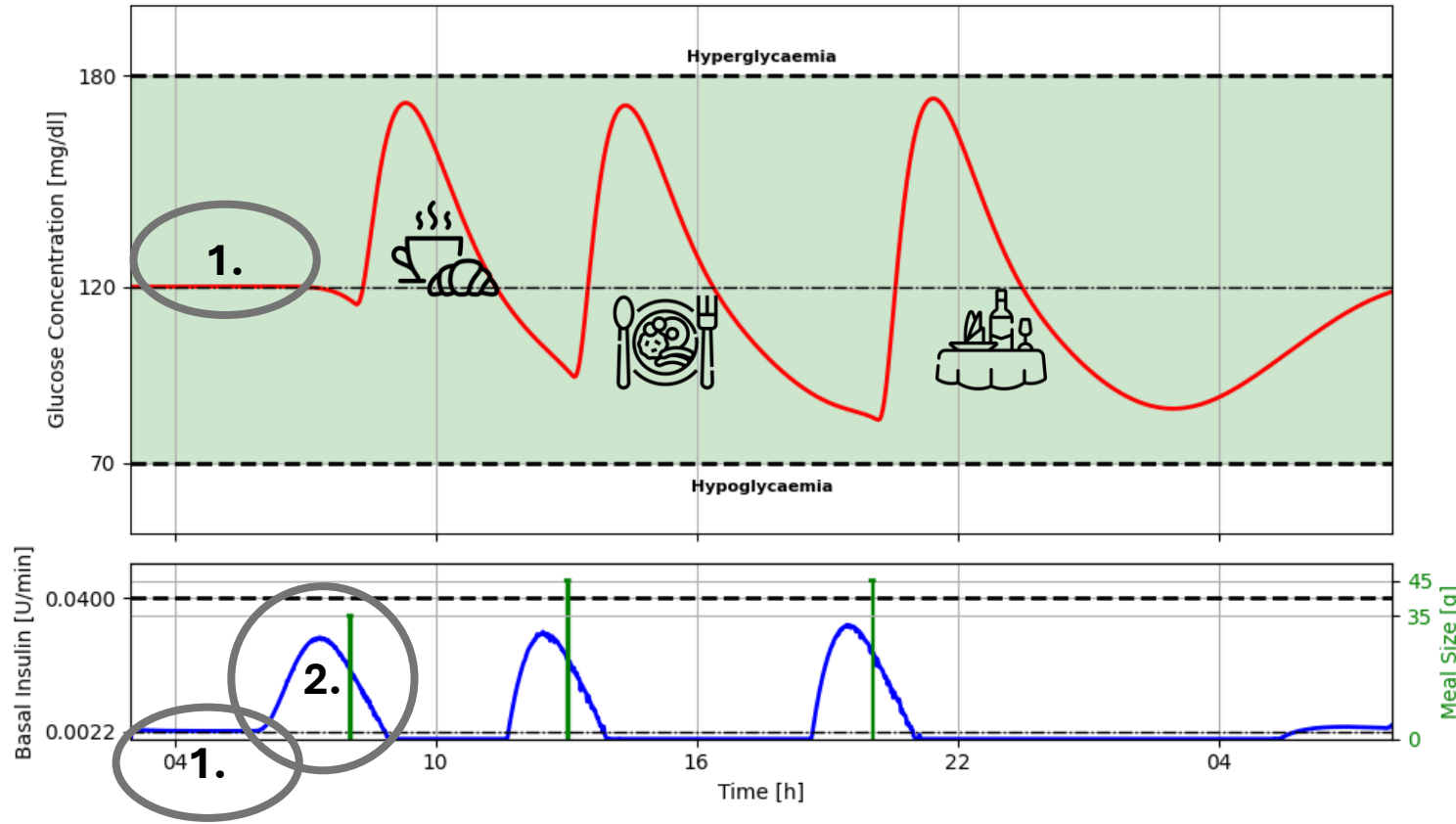
## Meal intakes

A severe failure scenario occurs when the patient consumes 60 grams of carbohydrates at breakfast and 30 grams at lunch and dinner.

## Limitations of current PID controller

- 1. Insulin effect delay:** The insulin's effect on the measured glucose concentration is delayed and not accounted for (the blue curve increases only after the green curve).
- 2. Constraints:** The controller cannot safely satisfy constraints under these conditions.
- 3. Risk of extremes:** Both hyperglycemia and hypoglycemia are possible outcomes.
- 4. Injection capacity:** The maximum injection capability of the insulin pump is not met, leading to insufficient insulin delivery in real life.

## Tracking MPC Controller



## Conclusion

- Meal prediction and anticipation of insulin injections counterbalance the increase in glucose concentration.
- Constraints are safely respected, ensuring optimal control and patient safety.
- Tuning parameters are tailored to every specific patient.

## Meal intakes

The default meal values are 35 grams for breakfast and 45 grams for lunch and dinner.

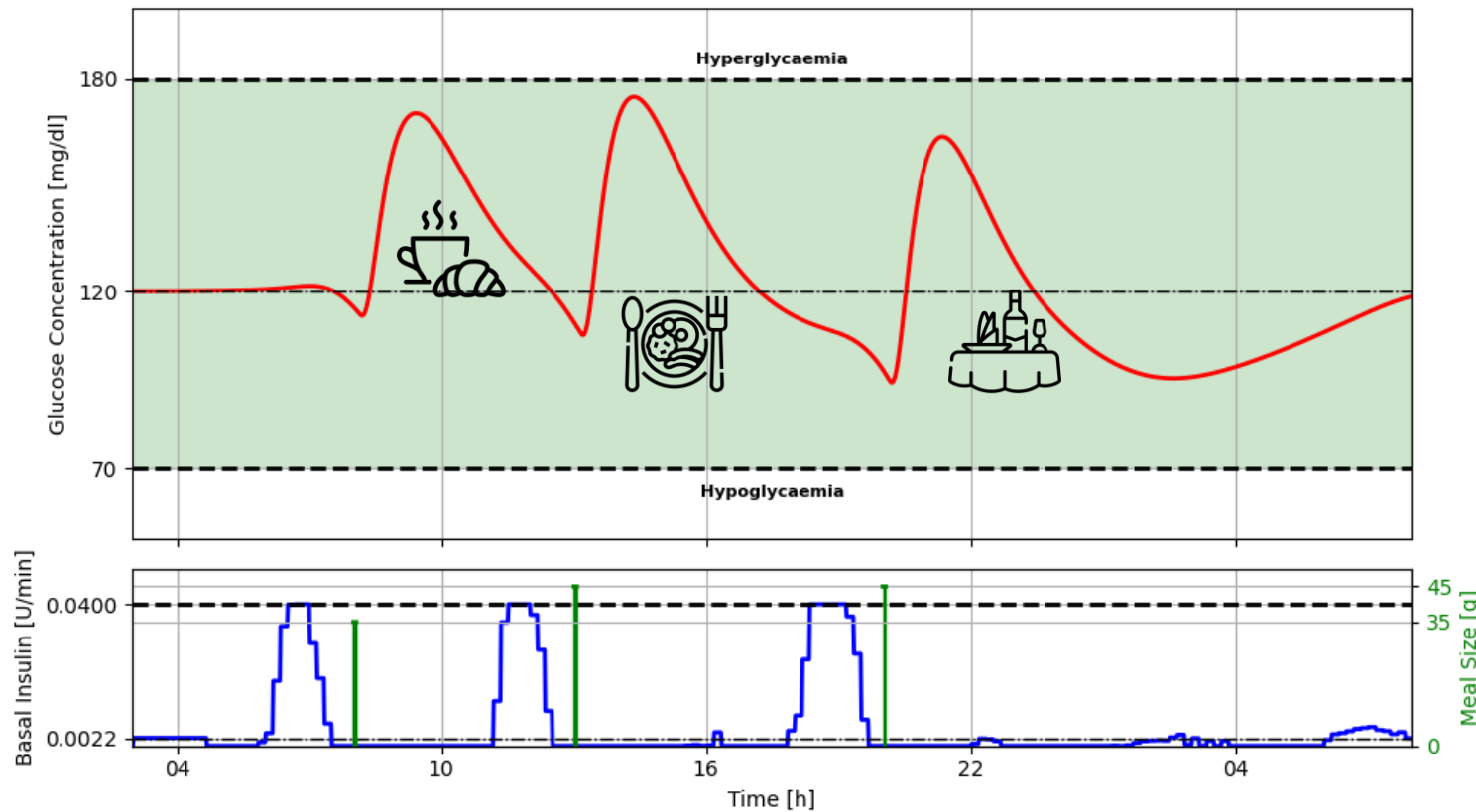
## Model-based controller

The recommended controller type is a tracking Model Predictive Control (MPC).

## Controller motivation

- 1. Tracking:** Ensures that the insulin and glucose level are maintained at the steady-state reference values of 0.0022 U/min and 120 mg/dL.
- 2. Predictive and receding horizon:** Anticipates meals and mitigates model mismatches, resulting in safer behavior.
- 3. Constraints:** Naturally integrates glucose and pump limits into the control formulation, enhancing safety and performance.

## DeePC Controller - Measuring $I_{sc1}$ and $G$



## Conclusion

- Meal prediction and anticipation of insulin injections counterbalance the increase in glucose concentration.
- Constraints are safely respected, ensuring optimal control and patient safety.
- Tuning parameters are tailored to every specific patient.

## Meal intakes

The default meal values are the same as before.

## Data-driven controller

The recommended controller type is a regularized tracking Data-enabled Predictive Control (DeePC).

## Controller motivation

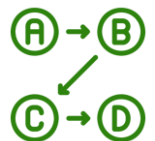
1. **Regularized:** Minimizes linear-nonlinear mismatches and ensures accurate prediction of the outputs.
2. **Tracking, constraints and receding horizon:** Like the MPC, it anticipates meals, consider constraints and mitigates mismatches, promoting safer behavior.
3. **Direct control:** Eliminates the need for an identification step by directly using collected patient data for prediction and control.



## Technical steps

Based on insights presented by Simone Del Favero, an expert in artificial pancreas controllers, the necessary steps for deployment are:

1. **Controller evaluation:** Use the ReplayBG simulator as alternative to animal testing to ensure that the controllers perform as expected for many different scenarios.
2. **Hardware integration:** Embed the controller into hardware equipped with a clear interface for flawless operation.
3. **Human trials:** Proceed with evaluations in controlled environments on human subjects to ensure safety and performance in real-world scenarios.



Specific for the tracking MPC



Specific for the regularized tracking DeePC

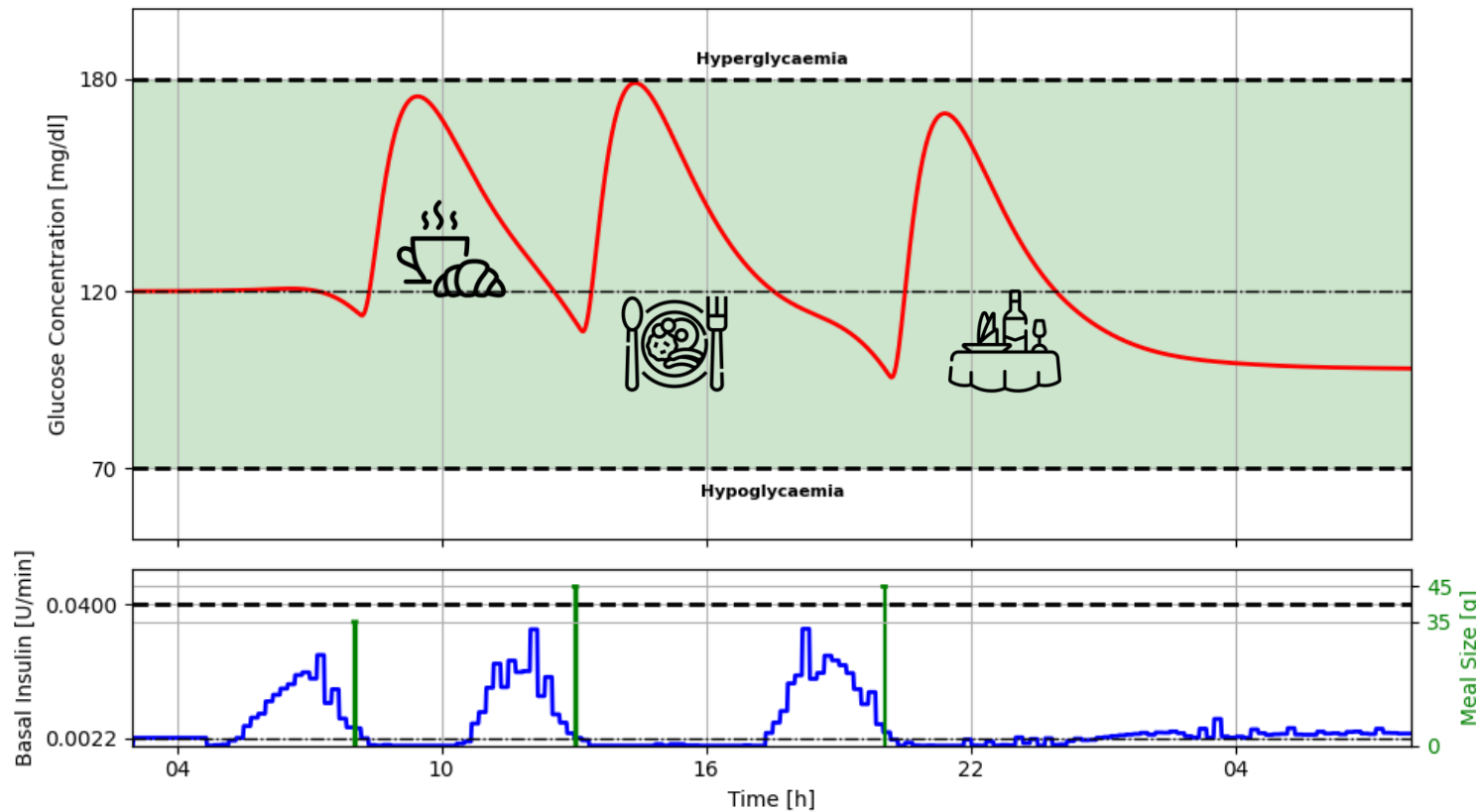
## Requirements

- **Safety:** The system must be safe at all times, meaning stability and robustness must be met.
- **Feasibility:** The controller must consistently produce feasible insulin actuation values.
- **Personalized tuning:** Parameters must be customized for each individual patient.
  - For MPC also ReplayBG model parameters.
  - For DeePC also high-quality and sufficient data.
- **Low power consumption:** The optimization calculation must be as efficient as possible.

## Trade-offs

- **Data vs. safety:** Balancing the need for data that covers a wide glucose concentration range with the requirement to safely obtain these values.
- **Precision/accuracy vs. computational effort**
  - **Long prediction horizon:** Using a longer prediction horizon for better future estimates.
  - **Low discretization step:** Employing a low discretization step for finer control.
  - **Frequent optimization:** Often solving optimization problems to maintain accuracy.

## DeePC Controller - Measuring noisy G



## Conclusion

- The presence of significant noise can cause tracking problems for the controller since it relies on data.
- This controller is more applicable in real-world scenarios as it only requires data from the patient.
- Tuning parameters are tailored to every specific patient.

## Requirements and merits

In real-world applications, only glucose concentration can be effectively measured. Therefore, the regularized tracking DeePC proposed earlier is modified to rely solely on this value. Additionally, this data is corrupted by noise, reflecting practical scenarios.

## Trade-offs

- **Data vs tuning:** The precision of the controller is highly dependent on the quality, quantity and glucose concentration range of the available data.
- **State information:** Relying on only one of the nine states of the model reduces the number of tuning parameters but increases the challenge of achieving a reliable and precise controller.



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