



BioPINNacle : Redefining Traditional Process Modelling in Bioethanol Production Through Physics Informed Neural Networks

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Introduction

- Bioethanol fermentation shows strongly coupled changes in substrate, biomass, and ethanol over time.
- Accurate modelling of these dynamics is essential for improving process efficiency.
- Conventional neural networks fit data but often ignore known fermentation kinetics.
- Physics-Informed Neural Networks (PINNs) combine data with kinetic equations for reliable and interpretable modelling.

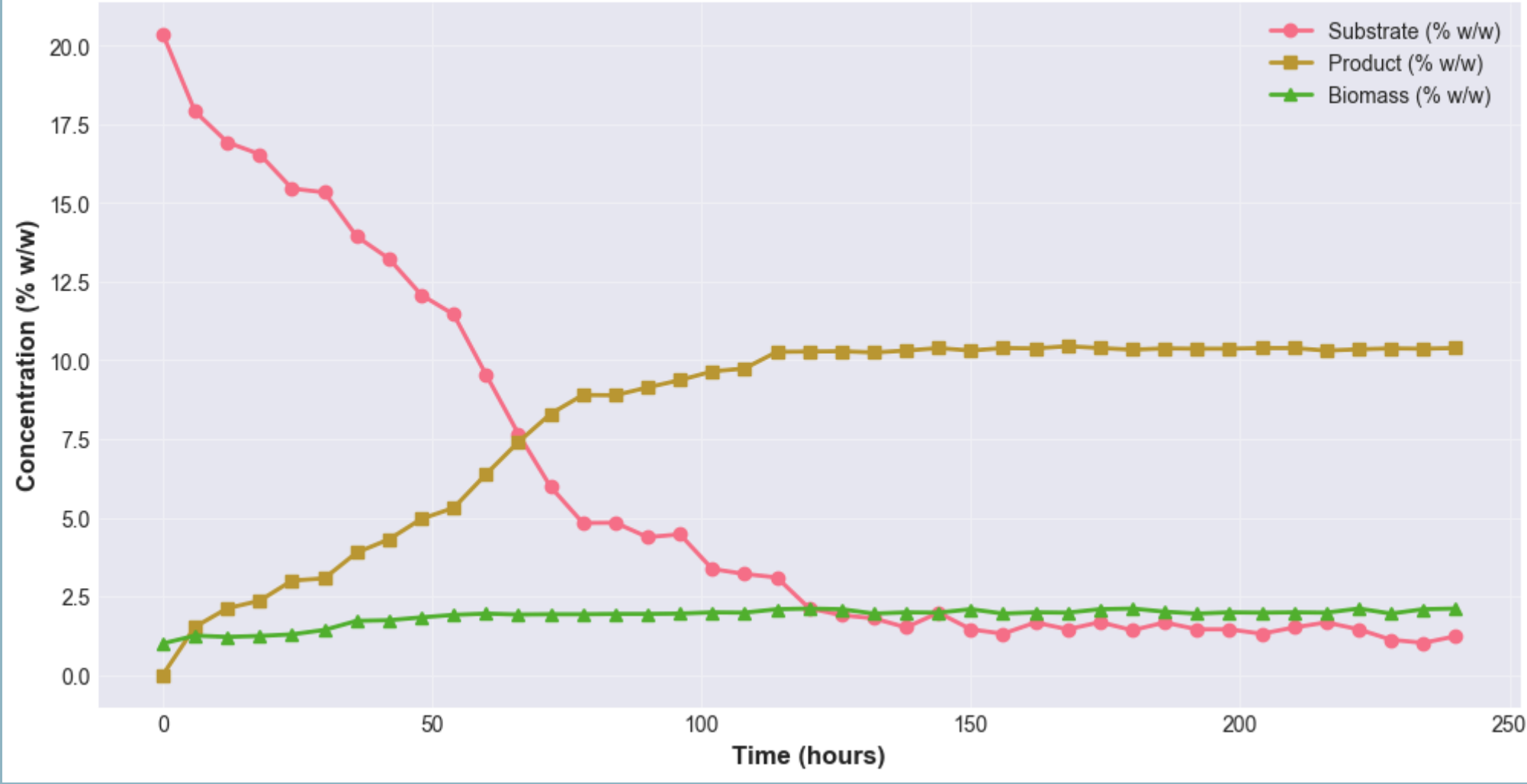
Problem Statement

- Bioethanol fermentation involves complex, time-dependent interactions between substrate consumption, biomass growth, and ethanol formation that are difficult to model accurately.
- Traditional data-driven models provide good curve fitting but often violate known fermentation kinetics and lack physical interpretability.
- There is a need for a modelling framework that can learn from limited experimental data while strictly adhering to governing biochemical laws.

References

- Raissi et al. (2019). J. Comput. Phys., 378, 686–707.
- Karniadakis et al. (2021). Nat. Rev. Phys., 3, 422–440.
- Asiedu et al. (2024). J. Eng. Appl. Sci., 71:195.

Cassava Fermentation Profile



- Substrate concentration decreases steadily over time due to microbial consumption.
- Ethanol concentration increases and eventually plateaus, indicating fermentation completion.
- Biomass exhibits initial growth followed by stabilization.
- Overall trends are consistent with Monod-based fermentation kinetics.

Governing Equations (Cassava)

- Biomass Growth

$$\frac{dX}{dt} = \mu(S, P) \cdot X$$

- Product Formation

$$\frac{dP}{dt} = q_P(S, P) \cdot X$$

- Substrate Consumption

$$\frac{dS}{dt} = -(1/Y_X S) \cdot \mu(S, P) \cdot X - (1/Y_P S) \cdot q_P(S, P) \cdot X$$

- Specific Growth Rate (Monod with Product Inhibition)

$$\mu(S, P) = \mu_{max} \cdot (S/(K_{sx} + S)) \cdot (1 - K_{ipx} \cdot P)$$

- Specific Product Formation Rate

$$q_P(S, P) = q_{Pmax} \cdot (S/(K_{sp} + S)) \cdot (1 - K_{ipp} \cdot P)$$

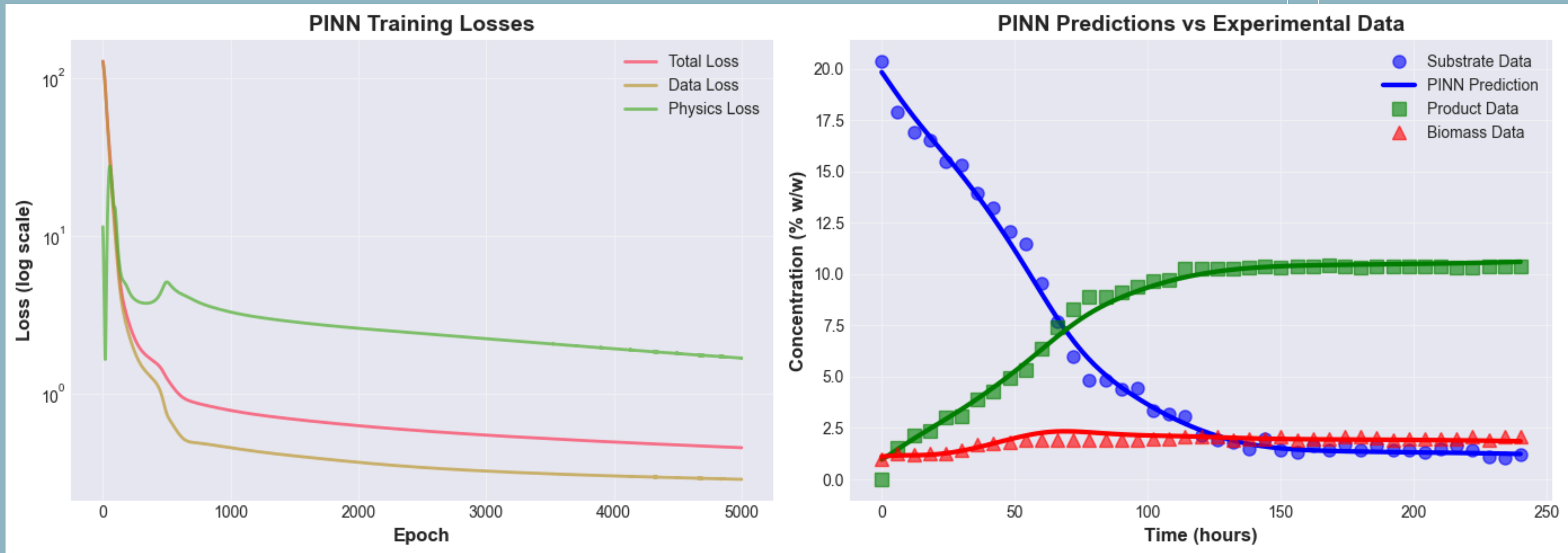
- Initial Conditions

$$X(0) = X_0 \quad S(0) = S_0 \quad P(0) = 0$$

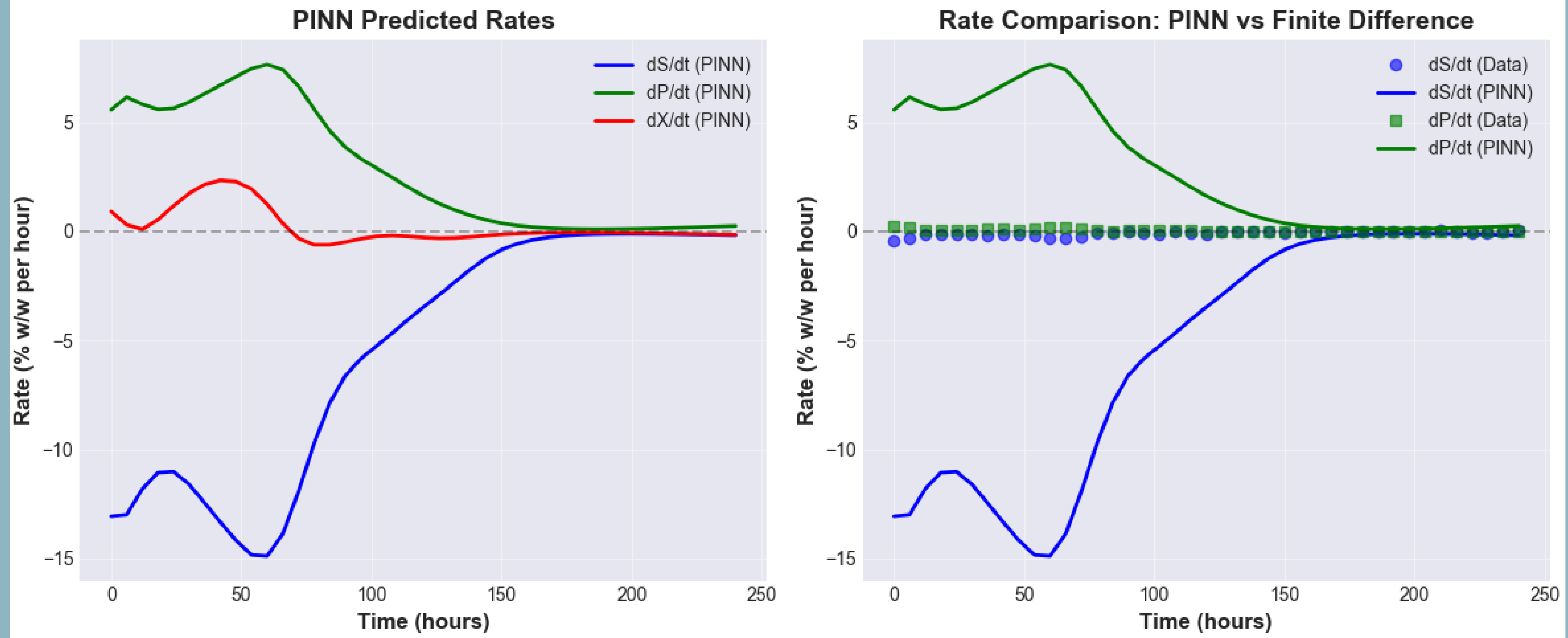
Learned Parameters

Parameter	Value
mu_max	0.841695
q_pmax	5.58997
K_sx	-0.929824
K_sp	2.356223
K_ipx	0.102616
K_ipp	0.06313
Y_xs	0.256961
Y_ps	0.54206
M_s	-0.116052
G_s	-0.106052

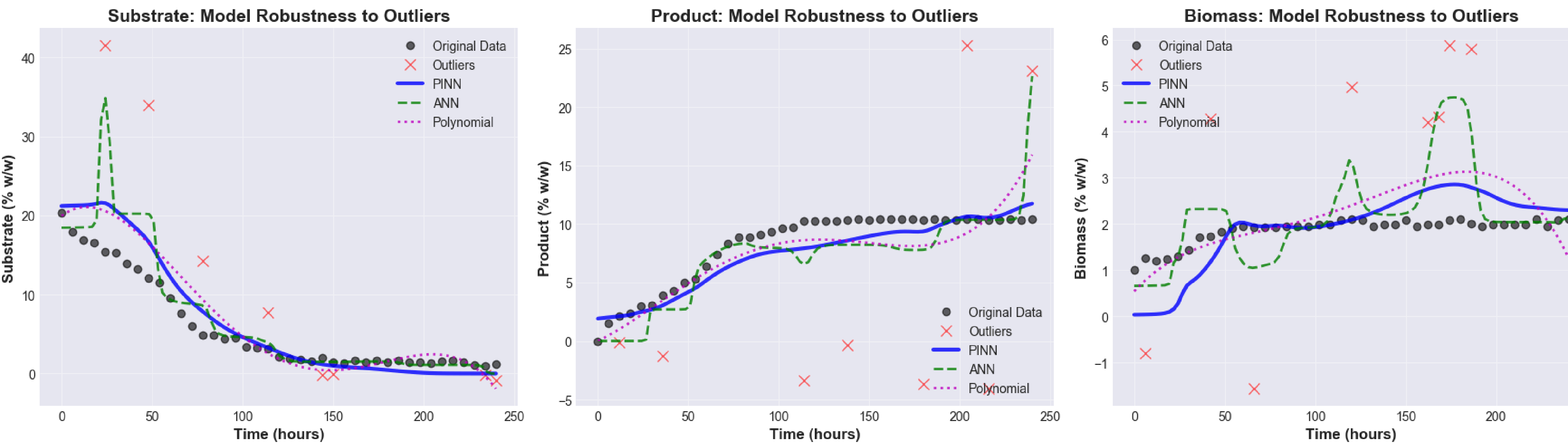
PINN Performance



Saliency Plot



Outlier Scenario



BioPINNacle App



The BioPINNacle app was created to provide a user interface and act as a digital twin for the bioethanol fermentation process.

R² Result For Outlier Scenario

Model	Substrate	Product	Biomass
PINN	0.8075	0.823	-3.3067
ANN	0.5748	0.3261	-9.6572
Polynomial	0.7903	0.6837	-3.6555