Construction of a Machine Learning tool based on Differential Equations integrated into a dynamic tree allowing biological data assimilation

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- Modeling of the digestive behaviour
- The goal: build Machine Learning algorithms based on the discretization of Differential Equations in order to predict biological responses and assimilate biological data.

Graph-oriented Model

• The main idea: Assimilate an animal to a graph in which nodes are virtual organs and edges represent fluxes within the organism.



Construction of a Machine Learning tool based on Different

Graph-oriented Model

• Each virtual organ carry a biological function, which aimed to reproduce at best its actual behaviour.



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Graph-oriented Model

• Each link is weighted with a parameter γ_i , on which indicates how inputs and outputs are actually dispatched, and to give more or less significance on links between organs.



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Graph-oriented Model

• Each node (organ) is given some memory, which are states. They can thus store and use several values, which can be part of their function definition.



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Graph-oriented Model

The process of such function is as follows:

- 1) Aggregate all the ingoing fluxes coming from other functions
- 2) Given those inputs, compute the function output for this time step.
- 3) Propagate the computed output to all outgoing edges.



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Global Optimization

Optimize the whole animal in order to find the parameters values that would make it fit at best the real data.

2 approaches:

1) Analytic approach: Analytically resolve the functions, by fixing some parameters each by each.

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2 approaches:

- 1) Analytic approach: Analytically resolve the functions, by fixing some parameters each by each.
- 2) Machine Learning approach: Optimization (Genetic optimization methods, Least square method...)
 ⇒ Fast to implement.

Applied case: Ross Chicken growth



ROSS 708

http://en.aviagen.com/brands/ross/products/ross-708

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Applied case: Ross Chicken growth



Some information on those animals are available, for each sex of the strains:

- -Daily consumption of food.
- -Daily growth (Weight).

ROSS 708

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- At first, we consider the growth(mass gained each day) .
- Flux to consider is the food consumption which becomes later a nutrients flux.

Recall of some growth functions

• Logistic:
$$W' = \mu_0 W \left(1 - \frac{W}{W_f} \right)$$

• Gompertz:
$$W' = DW \ln\left(\frac{W_f}{W}\right)$$

• Richards:
$$W' = DW\left(\frac{W_f^n - W^n}{nW_f^n}\right)$$

Assumptions



A1 Ingested food is stored in the crop before moving into the stomach.

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Assumptions



- A1 Ingested food is stored in the crop before moving into the stomach.
- A2 Some fluxes rates are constant between compartements.

Assumptions



A3 Existence of transmission delays in some compartments.

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Assumptions



- A3 Existence of transmission delays in some compartments.
- A4 There exists a time τ for which the SI empting dynamic starts.

Equations



Crop

$$egin{aligned} Q_{cp,d}' &= -\gamma_{cp} Q_{cp,d} \ Q_{cd,d}(0) &= S_d \end{aligned}$$

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Equations



Stomach

$$egin{aligned} Q_{st,d}' &= \gamma_{cp} \left(1 - \exp(-eta_1 t)
ight) Q_{cp,d} - \gamma_{st} Q_{st,d} \ Q_{st,d}(0) &= 0. \end{aligned}$$

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Equations



Small intestine

$$\begin{aligned} Q_{si,d}' &= \begin{cases} \gamma_{st} Q_{st,d}, \ t \leq \tau \\ \gamma_{st} Q_{st,d} - (\gamma_{abs} + \gamma_{si}) Q_{si,d}, \ t > \tau \\ Q_{si,d}(0) &= 0. \end{aligned}$$

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Equations



Large intestine

$$Q'_{Li,d} = \gamma_{si}Q_{Li,d} - \gamma_{Li}Q_{Li,d}$$
$$Q_{Li,d}(0) = 0.$$

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Storage and usage of nutrients

$$\begin{aligned} Q'_{w,d} &= \gamma_{abs} Q_{si,d} - \alpha Q_{w,d} \\ Q_{w,d}(0) &= Q_{w,d-1} \end{aligned}$$

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A6 Anabolism prevailing over catabolism (Pütter 1920).

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Formation of the daily weight

$$W'_d = \kappa(t)Q_{w,d} - \left(V_{max}rac{Q_{w,d}}{K+Q_{w,d}}
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 $W_d(0) = 0.$

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$$\begin{cases} \kappa' = -\mu\kappa \\ \mu' = \beta_2(\mu_{max} - \mu) \\ \Longrightarrow \kappa(t) = \kappa_0 \exp\left[-\mu_{max}\left(t - \frac{1 - \exp(-\beta_2 t)}{\beta_2}\right)\right] \end{cases}$$

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Final weight $W_f = W_0 + \sum_{d=1}^{N_d} W_d(t_f)$

 N_d : number of days.

- Observed data
 - Daily feed intake.
 - Daily weight measurement.
- Discretization: FDM(Forwards Euler method);
- fitting: γ_{cp} and τ fixed.

Estimate $\beta_1, \gamma_{st}, \gamma_{abs}, \gamma_{si}, \gamma_{Li}, \kappa_0, \mu_{max}, \beta_2, V_{max}, K$

Simulations



Figure: One day dynamic, $\beta_1 = 0.4, \gamma_{st} = 0.6, \gamma_{abs} = 0.6, \gamma_{si} = 0.023, \kappa_0 = 0.9, \mu_{max} = 0.01, \beta_2 = 0.01, V_{max} = 0.07, K = 26.8$



Figure: 2 days dynamic, $\beta_1 = 0.4, \gamma_{st} = 0.6, \gamma_{abs} = 0.6, \gamma_{si} = 0.023, \kappa_0 = 0.9, \mu_{max} = 0.01, \beta_2 = 0.01, V_{max} = 0.07, K = 26.8$

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Figure: $\beta_1 = 0.4$, $\gamma_{st} = 0.6$, $\gamma_{abs} = 0.6$, $\gamma_{si} = 0.023$, $\kappa_0 = 0.9$, $\mu_{max} = 0.01$, $\beta_2 = 0.01$, $V_{max} = 0.07$, K = 26.8

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Python code: Virtual chicken.

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- Mathematical description of biological functions.

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- Mathematical description of biological functions.
- Optimize different parameters.



Python code: Virtual chicken.

- Daily food consumption and it composition.
- Mathematical description of biological functions.
- Optimize different parameters.
- Predicts the daily growth.

Challenges and perspectives

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 - 1. No experience on Maching Learning.

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 - 1. Find needed DATA in order to build a better descriptive model.

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- Behaviour of a digestive track: Estimation of different fluxes rates. Important biological processes happening in different compartments.
- Perspectives:
 - 1. Find needed DATA in order to build a better descriptive model.
 - 2. Build a virtual chicken that allows fitting parameters using Machine Learning approach.

Références

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THANK YOU

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