

## ORIOLE: a web application for cleaning data from the walk-over-weighing device in livestock systems

Isabelle Sanchez, Eliel González García, Bénédicte Fontez, Bertrand Cloez

#### ▶ To cite this version:

Isabelle Sanchez, Eliel González García, Bénédicte Fontez, Bertrand Cloez. ORIOLE: a web application for cleaning data from the walk-over-weighing device in livestock systems. 74th Annual Meeting of the European Federation of Animal Science, Aug 2023, Lyon, France. Wageningen Academic Publishers, 29, pp.504, Book of Abstracts of the 74th Annual Meeting of the European Federation of Animal Science. 10.3920/978-90-8686-936-7. hal-04195753

## HAL Id: hal-04195753 https://hal.inrae.fr/hal-04195753v1

Submitted on 4 Sep 2023

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés. Book of Abstracts of the 74<sup>th</sup> Annual Meeting of the European Federation of Animal Science





Book of abstracts No. 29 (2023) Lyon, France 26 August – I September, 2023

# Book of Abstracts of the 74<sup>th</sup> Annual Meeting of the European Federation of Animal Science

Lyon, France, 26 August – 1 September, 2023



#### **EAAP Scientific Committee:**

F. Miglior L. Pinotti L. Boyle D. Kenny M. Lee M. De Marchi V.A.P. Cadavez S. Millet R. Evans L. Gasco M. Pastell G. Pollott (secretary) H. Spoolder (chair)



## Session 39

## ORIOLE: a web application for cleaning data from the walk-over-weighing device in livestock systems

*I. Sanchez*<sup>1</sup>, *E. González-García*<sup>2</sup>, *B. Fontez*<sup>1</sup> and *B. Cloez*<sup>1</sup> <sup>1</sup>INRAE, UMR MISTEA, 34060 Montpellier, France, <sup>2</sup>INRAE, UMR SELMET, 34060 Montpellier, France;

### eliel.gonzalez@inrae.fr

The use of the walk-over-weighing (**WoW**), which automatically records animal live weight (LW) in an automated, non-invasive manner, involves filtering the primary datasets produced by this technology. Removing outliers allows the correct data to be retained for a more consistent interpretation of individual daily physical activity progression. However, the standard methods used so far to perform this cleaning were impractical, time-consuming and required minimal mastery of the methods used. This limits the adoption of WoW by farmers and other end users. Our team previously developed a Kalman filter with impulse noised outliers algorithm for the automatic detection of outliers generated by the WoW (kfino; https://arxiv.org/abs/2208.00961). Once the kfino algorithm was tuned, the ORIOLE web-application was developed and deployed by our team (for OutlieRs detectIOn waLk wEighing; <u>https://oriole.sk8.inrae.fr/</u>). The Shiny library of the *R* software which enables to easily create user-friendly interactive web apps straight from R was used (https://shiny.rstudio.com/). Our web application allows users to import raw data measured from the WoW and through simple settings to perform outlier detection and weight prediction during the experiment. Descriptive statistics are then available such as number of daily weighing, evolution of weight per animal, evolution of the flock weight, 24 h kinetics of individuals. The web app is a dashboard composed of a menu of several subsets offering a user-friendly experience: (1) a 'Welcome' section; (2) the 'Genesis' of the technology and the web-app project; (3) the heart of the app with a section for the import and analysis of 'WoW data' and producing useful reports; (4) a 'How to' section documenting how to use the app. Users can analyse their data using full advantage of descriptive and statistics plots and download reports for communication and decision making.

p. 504



#### A web application for cleaning data from the walkover-weighing device in livestock systems



INRA

Isabelle Sanchez<sup>1</sup>, Eliel González-García<sup>2</sup>, Bénédicte Fontez<sup>3</sup>, Bertrand Cloez<sup>1</sup> INRAE, UMR MISTEA 34060 Montpellier FRANCE - 2 INRAE, UMR SELMET 34060 Montpellier FRANCE - 2 Institut Agro, UMR MISTEA 34060 Montpellier FRANCE



ORIOLE

#### **Motivating Example**

• Data from [1].

- Each black point is the measurement from the WoW at a certain moment is of the same individual. • Black points are  $Y_k$  in our model
  - Each gold point is a liveweight measured with a static scale.
  - They are our gold standard. For each measure k, we aim to recover: -the real weight  $X_k$
- 03 Day 2021 04 2021 05 -the status (OK/KO namely good/bad)

#### KFINO - an innovative algorithm

Kalman Filter for Impulse Noised Outliers (KFINO) aims to both filters the data and detects outliers [2]. It is based on the iteration of the following two steps: • Expectation step : KALMAN type Filter

- From  $\mathbb{E}[X_1], p, m$ , for all data time k
- \* Prediction of  $X_{k+1}$  from the model and  $X_k$
- \* Correction of  $X_{k+1}$  from  $Y_{k+1}$  depending if  $Z_{k+1} = \mathsf{OK}$  and  $Z_{k+1} = \mathsf{KO}$
- \* Prunning step: keep the more probable path

• Maximisation step: estimation of  $\mathbb{E}[X_1], p, m$  from explicit computation (no optimization)

⇒ KFINO is fully integrated in the ORIOLE web application to help users in the outlier detection and prediction.

A pre-traitement is done by removing obvious outliers that are points Out Of a Range (OOR) fixed by the user of ORIOLE.



- Outlier detection and weight prediction Descriptive statistics available (number of daily weighing, progress of liveweight at individual and flock level, 24h kinet-
- WoW), 24h kinetics of individuals...) • User-friendly experience with easy steps and an 'How to' section
- · Versioned web application, easily scal-
- $\bullet$  Up to 200 analyzed dynamics in  $\sim$  4 mins

Users can analyse their data using full advantage of descriptive and statistics plots and download reports for communication and decision making.

TechCare Project

Global Objective: Integrating innovative TECHnologies along the value Chain to improve small ruminant welfARE management

Original device: The Walk-over-weighing device is a prototype measuring automatically weights at **high frequency** in indoor or outdoor settings. The WoW 'allow to improve the animals welfare monitoring by limiting manipulation and providing tools for the farmer's management.

Issue: High-throughput data with a high ratio of outliers (usually more than half) due to the gregarious behavior of small ruminants.

Objective of this work: To conceive a friendly user tool for the automatic filtering of outliers produced by the WoW whatever the system

#### Model for impulse noised outliers

We assume that the latent weight  $(X_k)$  are modelled through an Ornstein Uhlenbeck process (first order autocorrelation)

 $X_k \mid X_{k-1} \sim \mathcal{N}\left(X_{k-1}e^{-a(t_k-t_{k-1})} + m(1-e^{-a(t_k-t_{k-1})}), \frac{\sigma_m^2}{2a}(1-e^{-2a(t_k-t_{k-1})})\right)$ 

with some physiological parameters  $m, a, \sigma_m$  and times of measurement  $(t_k)$ . Measurement k is well monitored ( $Z_k = OK$ ,) with probability p or badly monitored  $(Z_k = KO)$ . We assume that the measure of the WoW  $Y_k$  verifies  $V_{t} \mid (X_{t}, Z_{t} = \mathsf{KO}) \sim \mathcal{P}$ 

with some fixed law 
$$\mathcal{P}$$
 for the outliers, and

 $Y_k \mid (X_k, Z_k = \mathsf{OK}) \sim \mathcal{N}(X_k, \sigma_p).$ 

Initial weight and most of these parameters are unknown and individual.





[1] Leroux, E., Llach, I., Besche, G., Guyonneau, J. D., Montier, D., Bouquet, P. M., Sanchez, I., González-García, E. 2023. Evaluating a Walk-over-Weighing system for the automatic monitoring of growth in post-weaned Mérinos d'Artes ewelambs under Mediterranean grazing conditions. animal -open space (https://doi.org/10.1016/j.inopes.2022.100032). [2] Cloce, B., Fontez, B., Conzilez-García, E., Sanchez, I. 2022. Klama noise outoires: a robust sequential algorithm to filter data with a large number of outliers. https://arxiv.org/abs/2208.00961. hal-03740301.