

# **Bee Observer**

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### **Introduction - Bee Observer**







Bundesministerium für Bildung und Forschung





### Introduction - The honey bee

## The European Honeybee (*apis mellifera*)



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## Citizen Science – Lead Your Project

Science by nonprofessionals, citizens as data collectors or thinkers







## Citizen Science – Find Your Project



SimRa - Sicherheit im Radverkehr

Das Projekt Simita sammelt Daten über Farnadfahrten, um Statistiken über Beinaheunfalle und viel befahrere Streckenabschnitte aufstellen zu können. Mit diesen Daten können Gefahrenstellen erkannt und die Situation verbessert werden. Stima, zisch, Tichnis

metri -a



#### Deutsche Kolonialgeschichte – Wer war was?

entersententers extentionaligeschichte zu erforschen Durchforste Online- und Offine-Archive nach zeitgenössischen Quellen und führe die Daten in Wikidata zusammen. So wollen wir das Bild über dia damalige Zeit weiter vervollständigen. Anredfreshultang Geschichteft

metr ->



SAIN – Stäldtische Agrikultur Landwirtschaft in der Städt neu derikent Entwickle in Bonn und Oberhausen gemeinsam mit Stadtfarmer\*innen und Wissenschaftler\*innen die Stadien und Bereiche der Nahrungsproduktion weiter und finde neue Ideen für Lebensmittel aus der Stadt für die Stadt Dreitrung Landmütung, Mexen

mehr ->

mater in



https://www.buergerschaffenwissen.de/projekte



### Citizen Science – Challenges and Chances

Find more relevant research questions

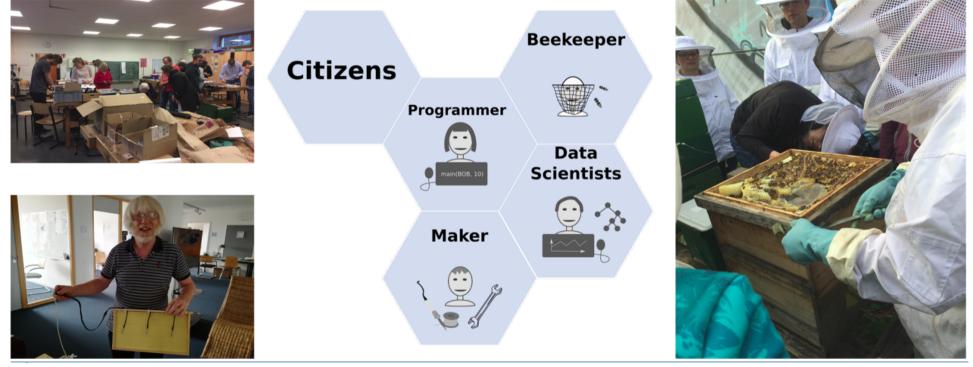
Make best use of available knowledge and experience

- Collect more data from different locations
- x Mostly no experimental setup
- × How to pay citizens for their work?

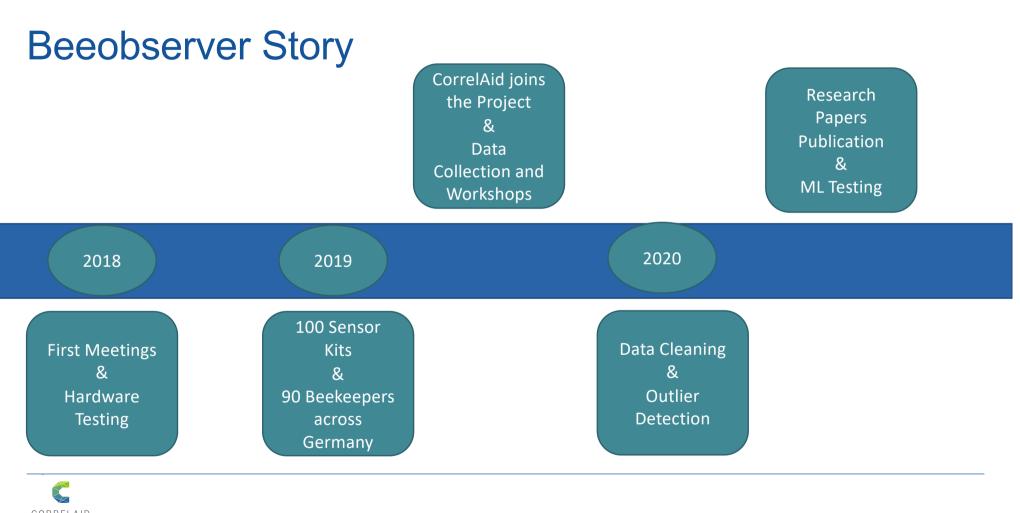




### Citizen Science – Beeobserver







CORRELAID

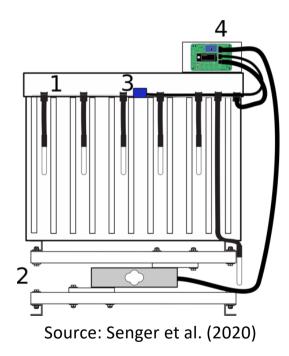
### Bee Observer - Sensor Data

Data collection since February 2019

Sensor-kits:

- 1. six thermometers (5 in, 1 out)
- 2. load-cell (scale)
- 3. humidity sensor
- 4. processing unit

Data stored in *influxdb* 





### **Bee Observer - Inspections App**

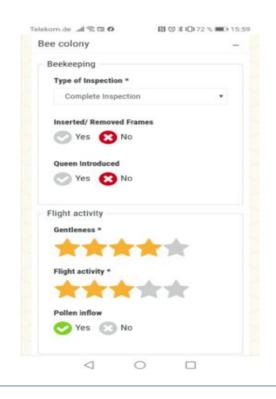
#### Sensor data is complemented by

#### **Beekeeper observations**

- mix of Boolean, scores, free text, categorical and numeric data
- most values voluntary

#### **Apiary metadata**

- type of hive, race of queen
- geolocation (voluntary)

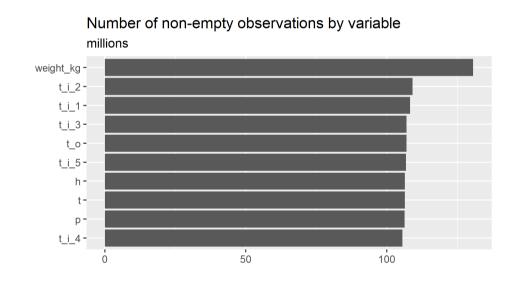




## Data Profiling - Sensor Data

#### hundreds of millions of observations from 129 beehives BUT

- variation across variables
- regular gaps
- some sensor kits do not have all sensors
- values outside the technical sensor range



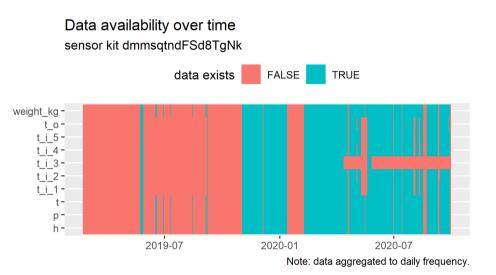
Note: Data until 30 September 2020.



## **Data Profiling**

The main issues in terms of data quality are:

- no data at all (interruption of electricity or WiFi)
- incomplete observations (single components fail)



Heatmap of data availability for an example beehive

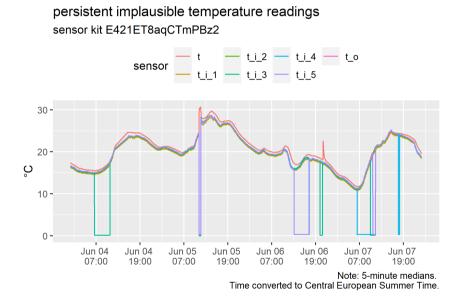


### Data profiling - Data Correctness

The other problem (less frequent) is obviously incorrect data:

- values outside the technical sensor range
- values within the range, but displaying implausible fluctuation (and/or lack thereof)

### Some cleaning needed!



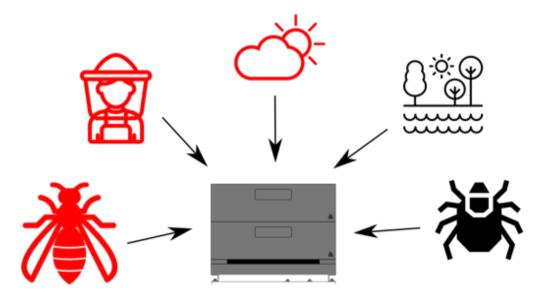
Example for apparent issues with thermometers



### **Local Outlier Detection**

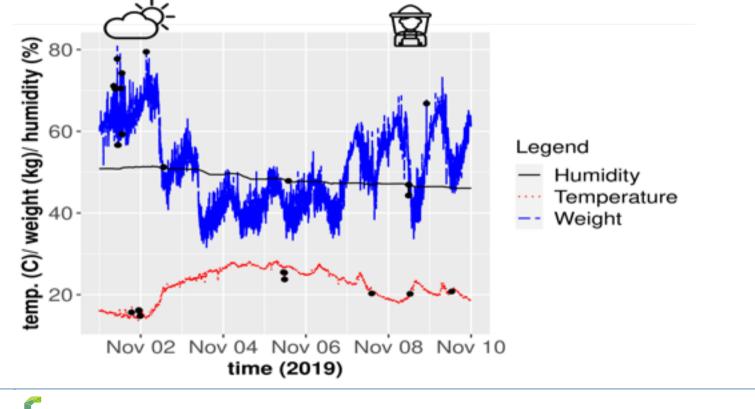
Data Modelling & Sensor Measurements Predictions

Difference between Prediction and the Actual Value





### **Local Outlier Detection**





### Analytical Use Case for Supervised Learning -Swarm Detection

When colonies swarm, the beekeeper has around one day to react. Otherwise the swarmed part of the colony may die or at least be lost to the beekeeper. Also the "old" part of the colony may be weakened.

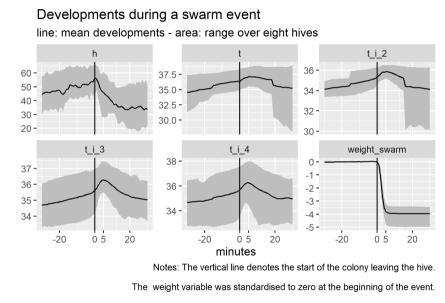
Swarms can be characterised by typical developments!



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### What happens during a swarm?



Developments in key variables during a swarm



### How do we model?

Non-linear model based on key variables, their growth rates and lags, time of day.

- small sample with missing data -> dimension reduction to maximum thermometer value, but also multiple models on subsets of variables
- unbalanced sample (few observed swarms): up-sampling training set with SMOTE

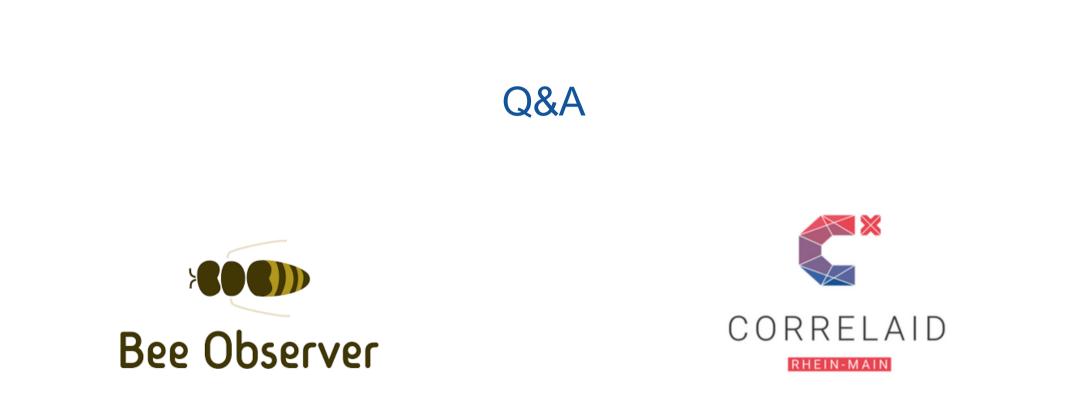
Random Forest works quite well

based on 10-fold cross-validation on the training set number of active components 2 3 no humidity unit 88% thermometers no scale 84% specificity no thermometer humidity unit 76% scale 97.2% 96.9% 96.6% 97.5% accuracy

Model performance by number of used sensors

### Performance metrics of RF models using subsets of sensor components







### **Introduction - Precision beekeeping**

The colony is a complex superorganism

- complex organisation
- division of labour
- developments are hard to observe: inspections are stressful, in winter not possible at all

Precision beekeeping

- install **sensors within beehive** for continuous monitoring
- weight, temperature, humidity, gas concentration, video, audio
- first works using simple algorithmic rules
- later first steps using ML approaches
- most based on small samples



### Introduction - The honey bee

## The European Honeybee (*apis mellifera*)

- important for pollination
- producer of honey
- increasing number of colonies in DE
- but increasing cases of colony losses over the winter



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### Outline

- Topical context Bees and beekeeping
- Bee Observer as a Citizen Science project
- The data
- ML use cases

