# Machine learning models for diagnostics and risk analysis

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**bTB** Diagnostics

• Goal:

- Augment the results of the Single Intradermal Comparative Cervical Tuberculin (SICCT) test with surrounding epidemiological risk factors:
  - to improve herd-level test sensitivity,
  - thus improve early detection and reduce onward transmission.
- Outcomes:
  - Integration of a large number of risk factors, including:
    - Farm characteristics, movements, testing, vet practice, tuberculin batch, ...
  - Herd-level prediction of bTB breakdowns using a machine learning model.
  - Risk factor breakdown using feature importance
    - and deeper analysis.

#### Foundation

• Initial idea from:

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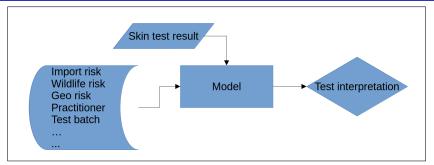
Using machine learning improves predictions of herdlevel bovine tuberculosis breakdowns in Great Britain

K. Stański 🖾, S. Lycett, T. Porphyre & B. M. de C. Bronsvoort

Scientific Reports 11, Article number: 2208 (2021) Cite this article

- Improved by:
  - Fewer (streamlined) risk factors, more historical data
  - Histogram-based Gradient Boosted Tree model:
    - Tuned and cross-validated model performance
    - Simplifying the treatment of missing data
  - Including new data from UK Farmcare (Vet practice and tuberculin batch).

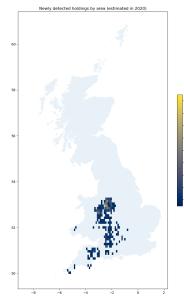
## Model and Data

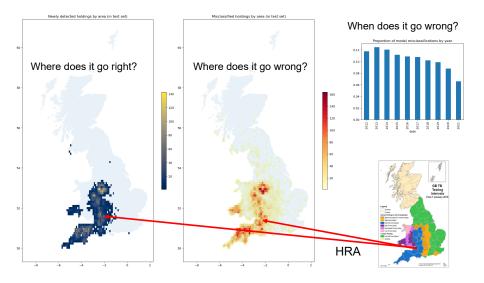


- Data extracted from APHA bTB surveillance database (SAM) and the Cattle Tracing System (CTS) database.
  - Every recorded SICCT test event between January 2012 and September 2021 (1.3m records)
  - Metadata on herd, location, movements, ...
- Data from UKFarmcare on vet and tuberculin batch for some tests:
  - 400 vet practices, covering 120k tests
  - 650 tuberculin batches, covering 57k tests

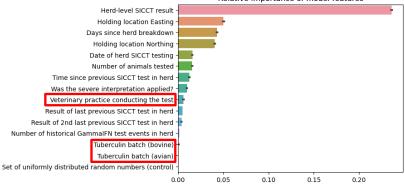
#### Results

- Model goal is to predict a new breakdown within 90 days of testing.
- Balance of sensitivity vs. specificity can be tuned in the model.
- Here we take the same herd-level specificity as the skin test alone, maximising sensitivity.
- Increase in herd-level sensitivity 12%.
- Over one year (2020) we find around 400 negative herd-level tests, that went on to have breakdowns, but were identified as positive by the model.





## Analysis - Risk factor importance

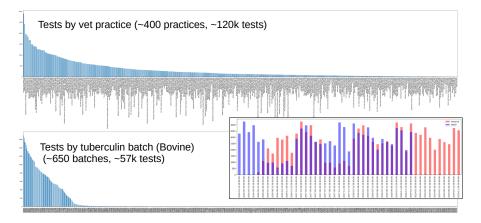


Relative importance of model features

- Risk factors ranked by permutation-based importance testing.
- Bar length shows accuracy reduction when factor removed from model.

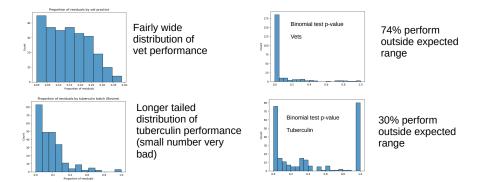
#### Vet data – Coverage

- Can we further explain risk factor importance of vet data?
- Data coverage may be reducing effectiveness:



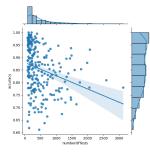
## Vet data – Vet/tuberculin accuracy

#### • What can we say about vet / tuberculin performance?



## Vet data - Performance by practice type

#### • Which practices perform worse?



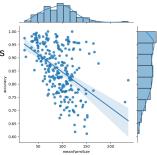
Size of vet practice by number of tests conducted has a weak correlation

r = -.27

Mean size of herd dealt with by the vet is correlated

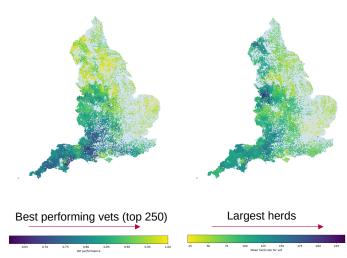
r = -.48

Vets that deal with larger herds perform worse



## Vet data – Performance by practice type

• Geographically?

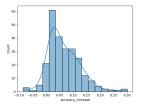


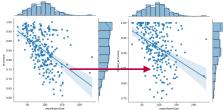
## What is the point of all this?

#### How does the model help?

Model augmented testing improves accuracy for most vets.

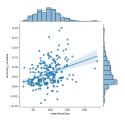
Some up to 30%:





Model reduces accuracy corelation with mean farm size from r=-.48 to r=-.38

> It increases performance with larger herds:



- UK Farmcare to conduct further analysis on vet results
- Possible extension to use skin test measurements instead of binary results
- How could this be used in the field?
  - App interface to model for vets to use at testing?
  - A guide to re-testing high-risk herds?
- Other situations: BVD, Johne's.