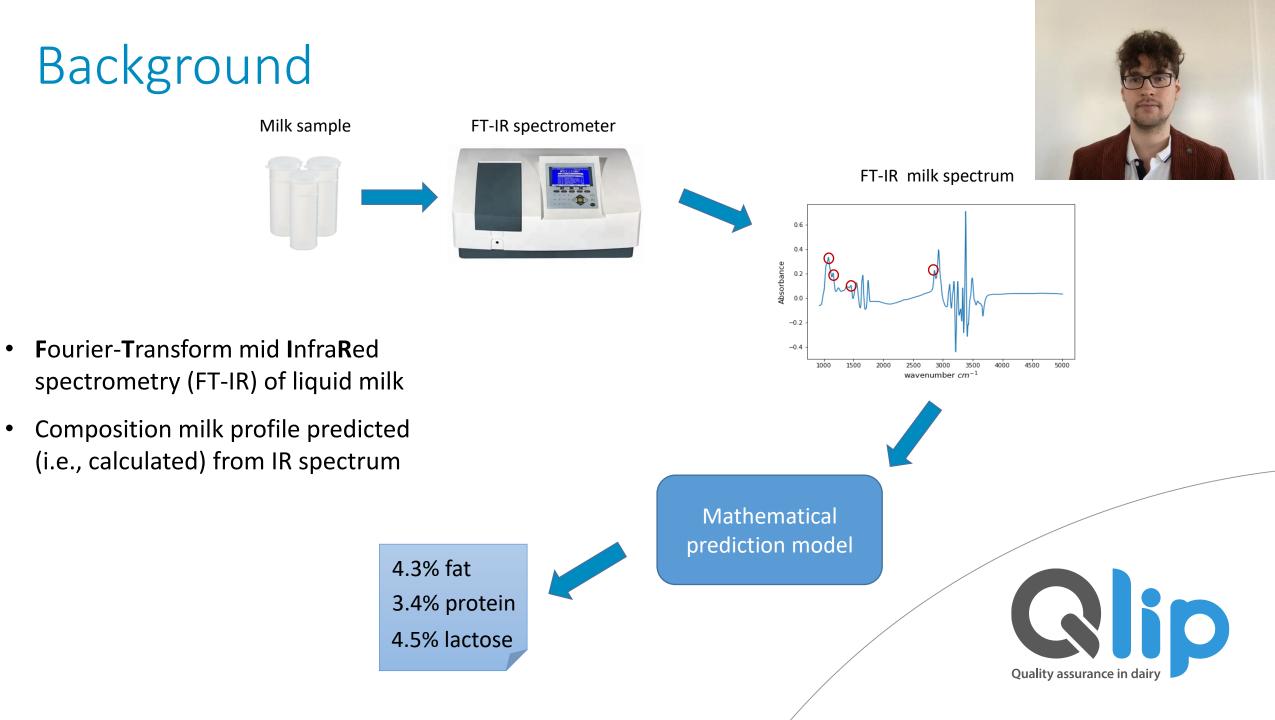
#### proficient · reliable · analytical · en Atypical Spectra Screening: Applications for Monitoring Infrared Instruments and Model Predictions

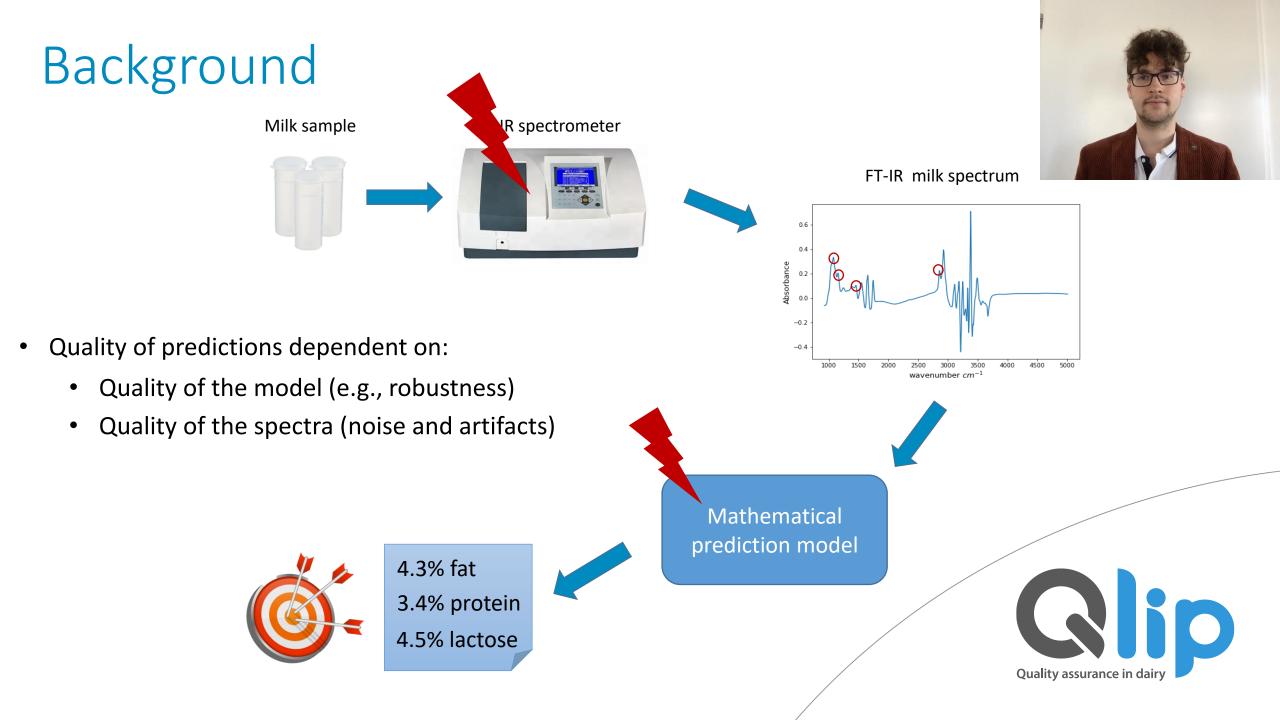




Lukas Spieβ, PhD Data scientist, Innovation and Business Development Qlip ICAR 2021







- Each spectrum results from **measurement process** performed under...
  - specific conditions
  - using a specific IR instrument



- Each spectrum results from **measurement process** performed under...
  - specific conditions
  - using a specific IR instrument
- Changes in measurement process (e.g., instabilities in the IR instrument) can lead to changes in the spectra

#### • Problematic:

- ...if this happens in spectral regions containing chemical information used by prediction models
- ...because the instrument is not working properly







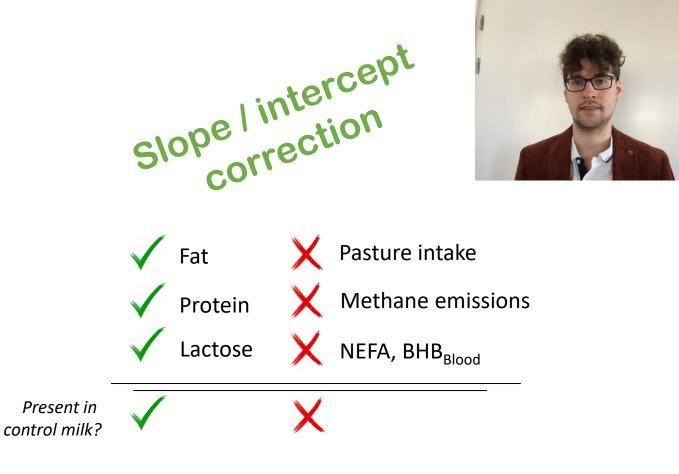








• Limited use-case for slope / bias corrections





- Limited use-case for slope / bias corrections ٠
- We need a general approach for monitoring ٠ the quality of our data (spectra, predictions)

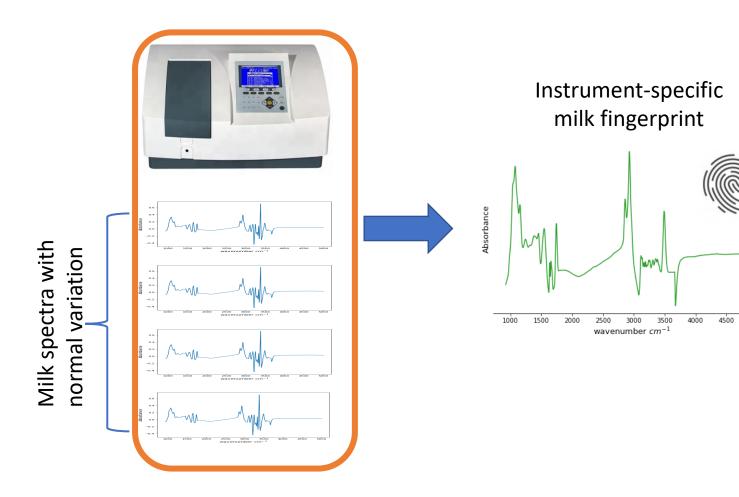
Present in

Slope | intercept correction Y Pasture intake 🗸 Fat **V** Protein **X** Methane emissions ✓ Lactose X NEFA, BHB control milk?

Quality assurance in dairy

- Limited use-case for slope / bias corrections
- We need a general approach for monitoring the quality of our data (spectra, predictions)
- Monitor spectra for systematic deviations due changes in the measurement process (i.e., instrument instabilities).
  - Use mathematical models originally developed to screen for non-specific adulteration in milk samples

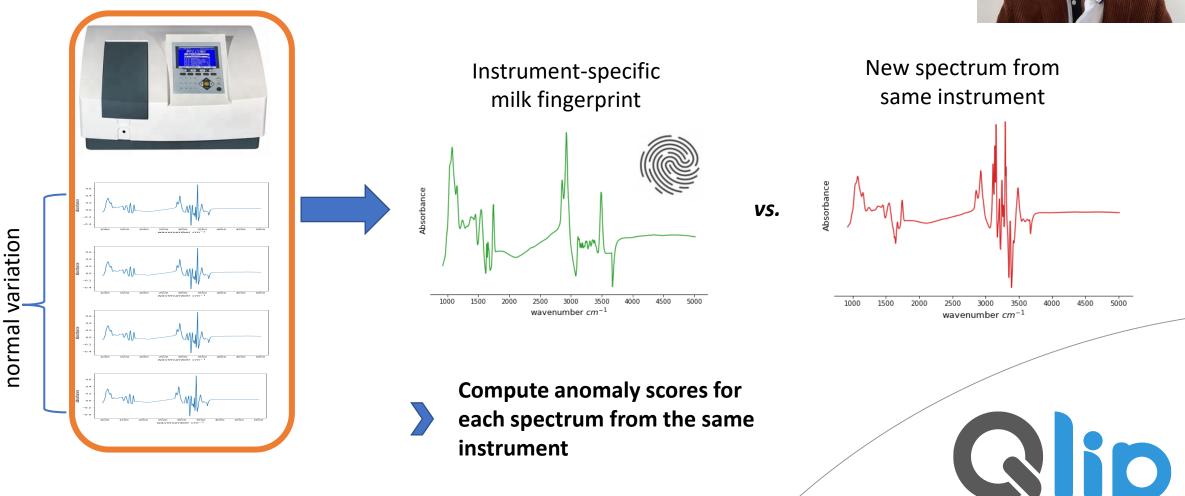
Slope | intercept correction Pasture intake 🗸 Fat V Protein X Methane emissions Lactose X NEFA, BHB Present in control milk? Quality assurance in dairy



Quality assurance in dairy

5000

Milk spectra with



Quality assurance in dairy



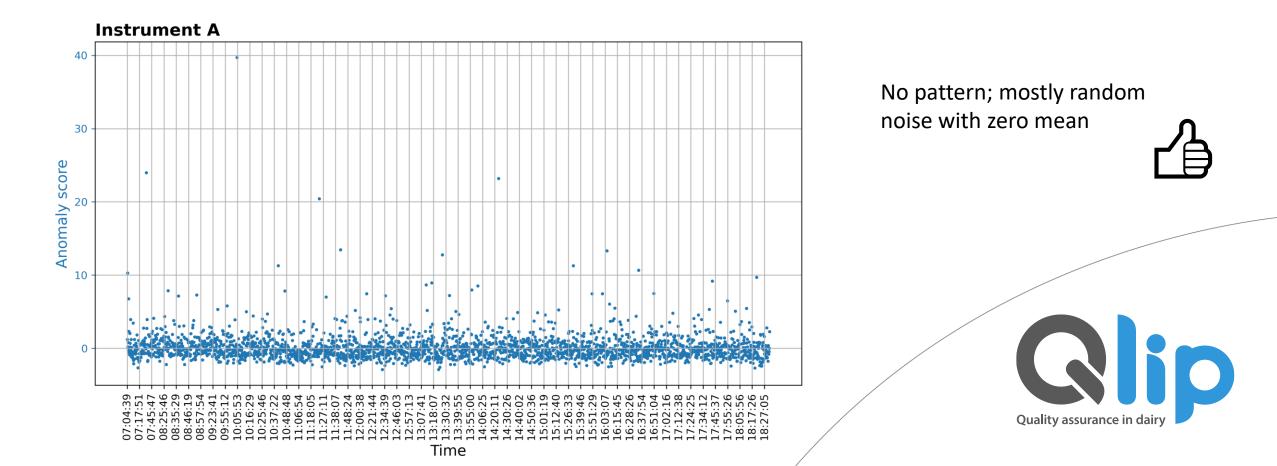




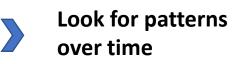




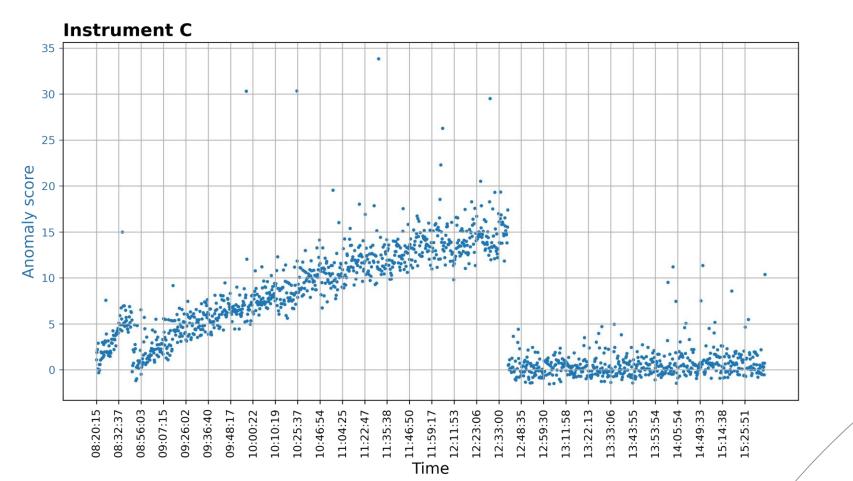




Compute anomaly scores for each spectrum from the same instrument



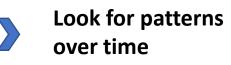




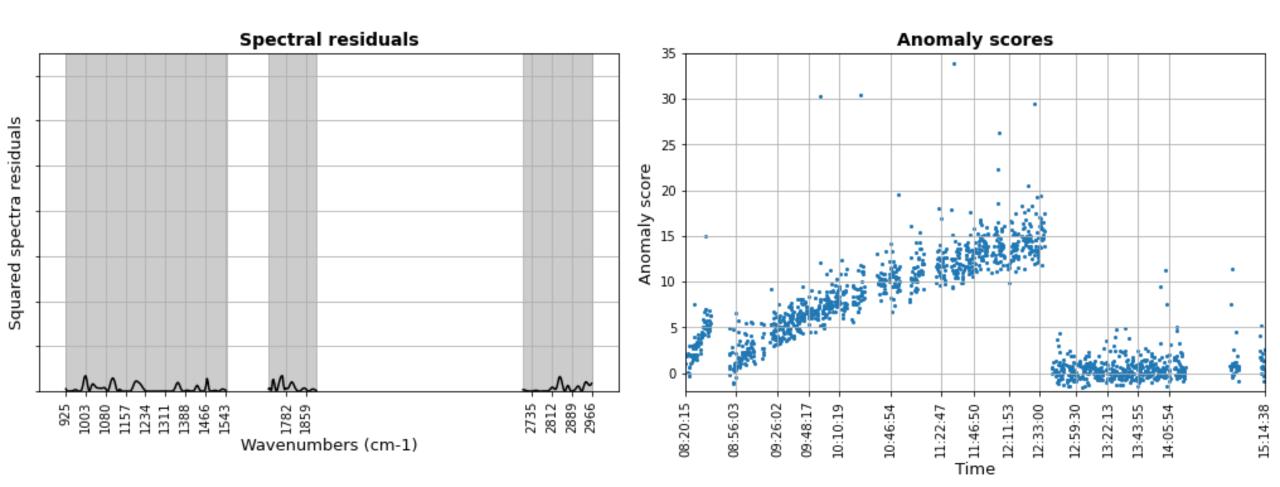


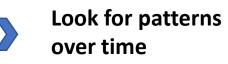
Gradual increase of anomaly scores



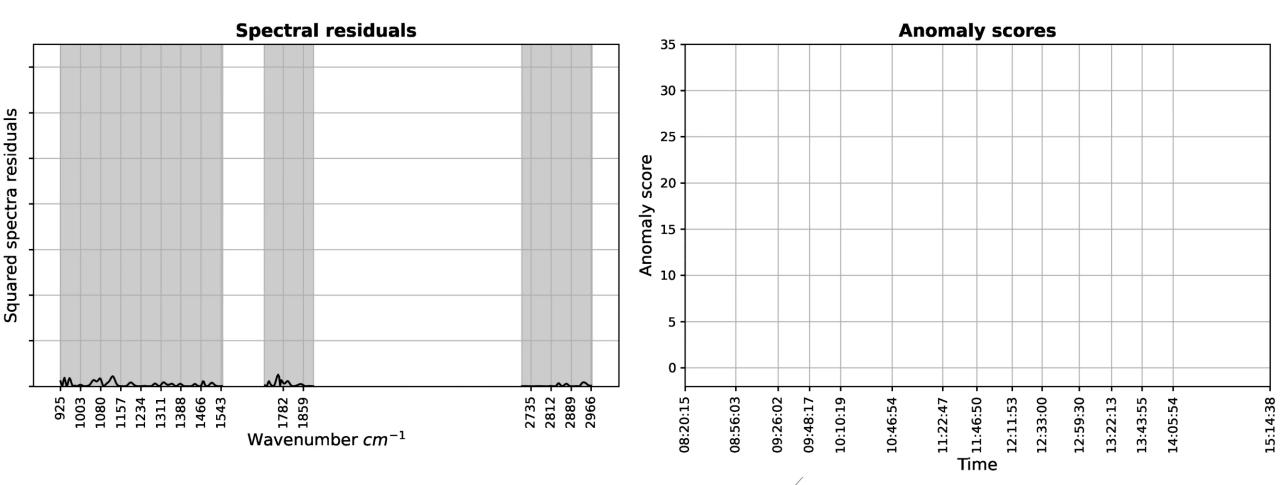


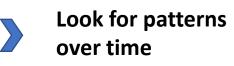


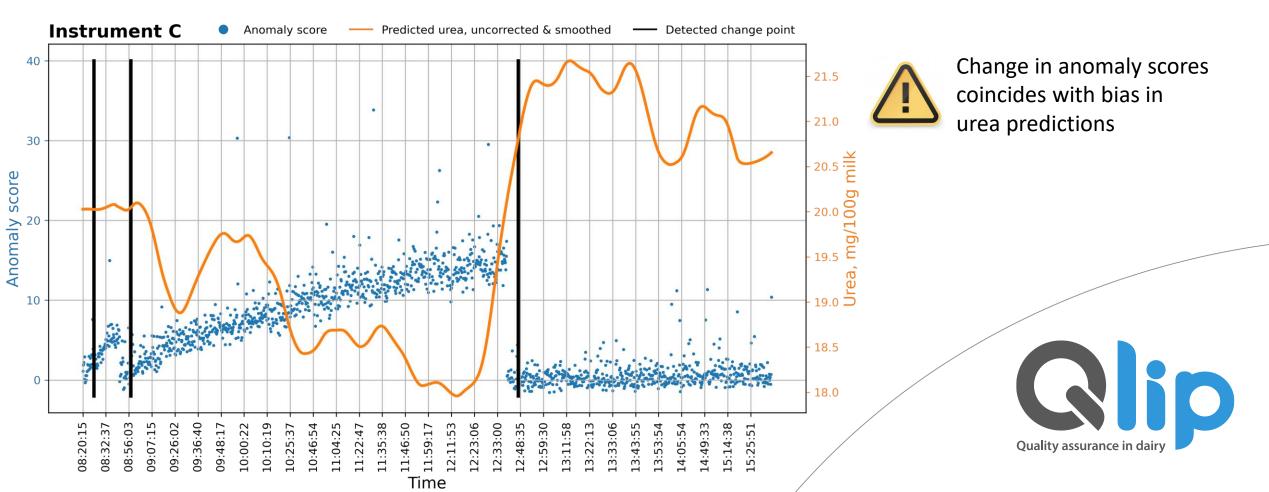


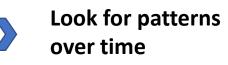




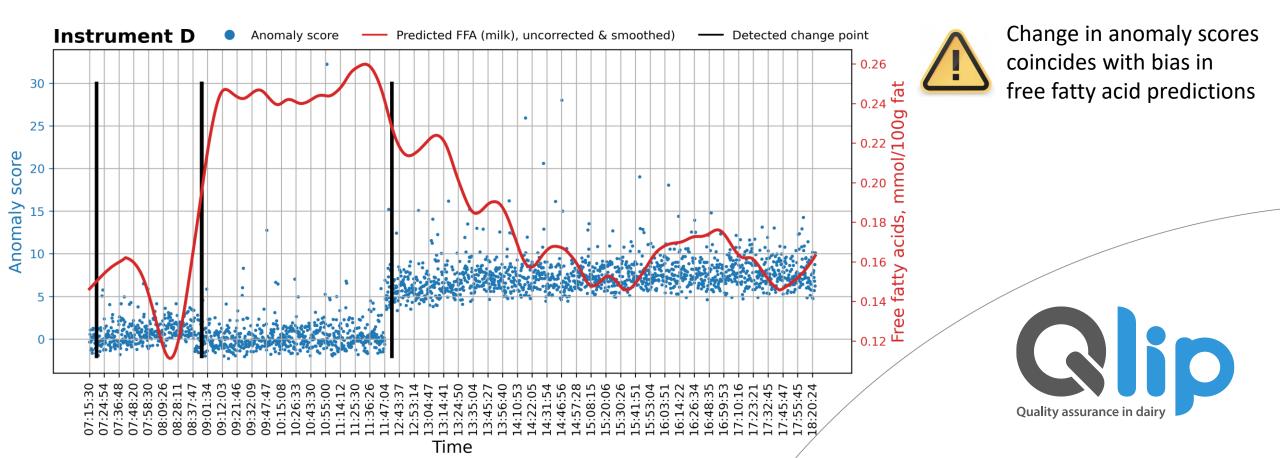




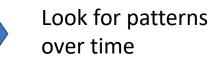








Compute anomaly scores for each spectrum from the same instrument



Detect systematic deviations automatically and real-time



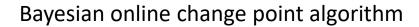


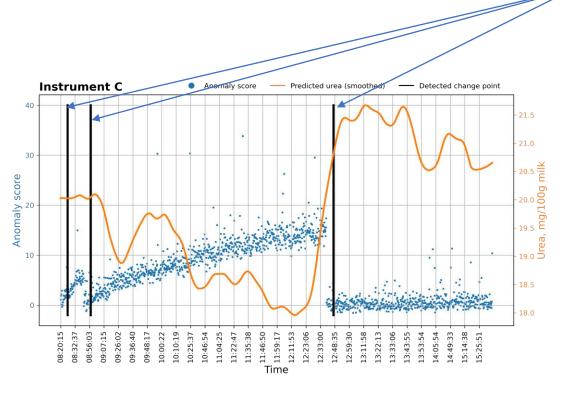
Compute anomaly scores foreach spectrum from the same instrument

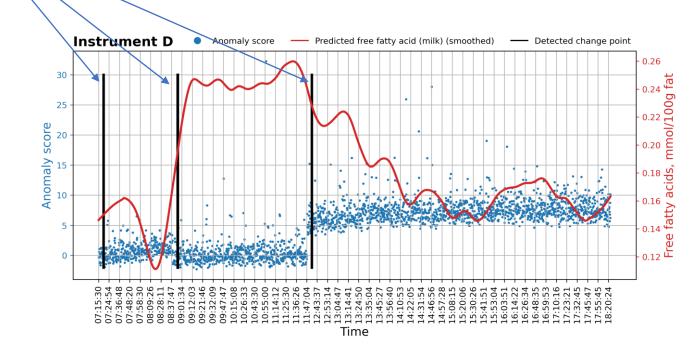


Look for patterns over time

Detect systematic deviations automatically and real-time









Compute anomaly scores for each spectrum from the same instrument



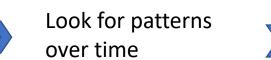
Look for patterns over time

On a scale of months to years

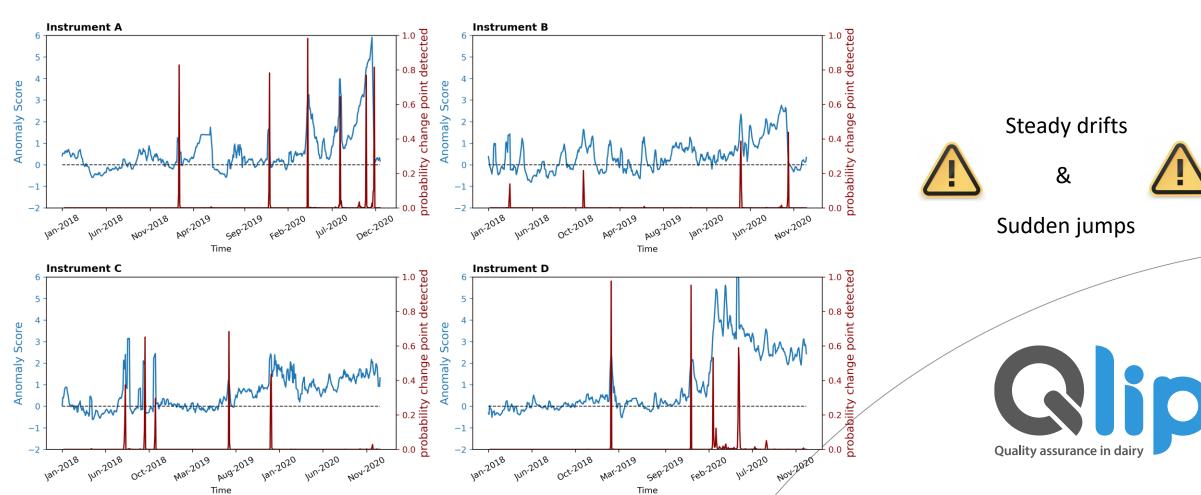




Compute anomaly scores for each spectrum from the same instrument



On a scale of months to years





Compute anomaly scores for each spectrum from the same instrument

Identify episodes with instabilities

Find patterns in spectral residuals





Compute anomaly scores for each spectrum from the same instrument

Identify episodes with instabilities

Find patterns in spectral residuals



Collected > 20.000 spectra from 38 episodes with instabilities identified across four instruments within a period of three years (> 5.9 million spectra)



Compute anomaly scores for each spectrum from the same instrument

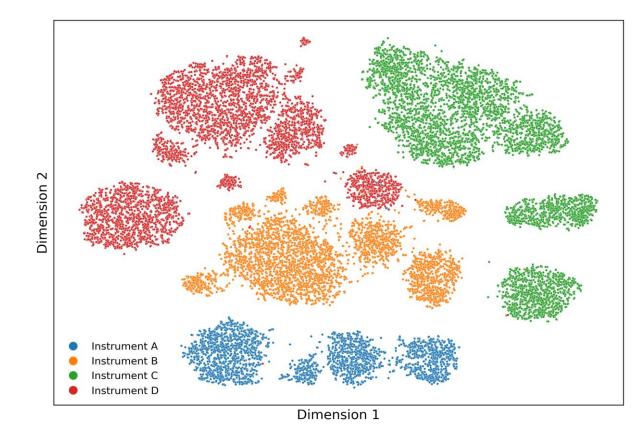
Identify episodes with instabilities

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PCA + *t*-distributed stochastic neighbor embedding of spectral residuals



Compute anomaly scores for each spectrum from the same instrument

Identify episodes with instabilities

Find patterns in spectral residuals

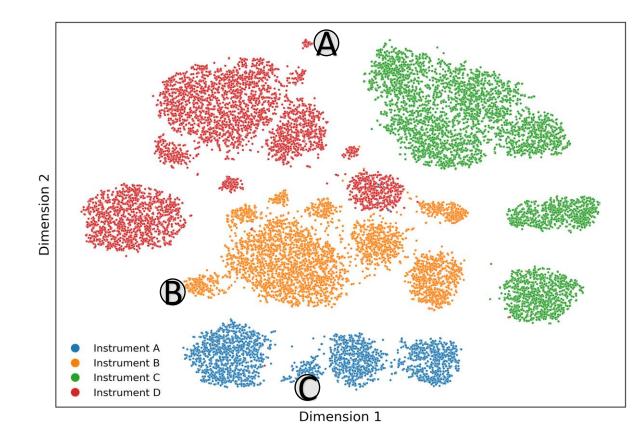


Collected > 20.000 spectra from 38 episodes with instabilities identified across four instruments within a period of three years (> 5.9 million spectra)

PCA + *t*-distributed stochastic neighbor embedding of spectral residuals

Many instabilities uniquely occur for short period





Compute anomaly scores for each spectrum from the same instrument

Identify episodes with instabilities

3

Find patterns in spectral residuals



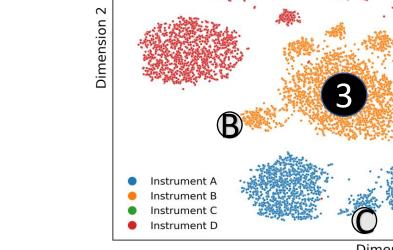
Collected > 20.000 spectra from 38 episodes with instabilities identified across four instruments within a period of three years (> 5.9 million spectra)

PCA + *t*-distributed stochastic neighbor embedding of spectral residuals

Many instabilities uniquely occur for short period



Some instabilities appear more often









- Our approach can be used as tool for:
  - Monitoring the validity of our model predictions
  - Monitoring the status of our instruments
- Does not rely on expensive chemical reference analyses or control milk samples.
  - Instead, it is based on the spectra that are obtained by routine IR analysis of milk samples
- Can be developed with the tools that already exist
- Can be developed and used by every laboratory





# Thank you for your attention!

