

# PLS2 regression as a tool for selection of optimal analytical modality

- *a closer look at beer brewing process analysis with alternative spectroscopic data types*

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## Aim of the study

- To compare the performance of three different matured spectroscopic techniques for beer brewing process monitoring (complex biochemical matrix)
- To investigate inter-relationships between spectral data types using the PLS2 algorithm
- To quantify the amount of unique information contained in each spectral data type based on bilateral PLS2 (e.g. NIR vs. Raman and vice versa)
- To propose a dedicated PLS2 approach for selection of optimal analytical modality for any given chemical matrix (biochemistry, environmental engineering, bio energy, chemical engineering)

## Bilateral PLS2 approach (*in development*)

- Clean-up spectral data for gross outliers and other unrepresentative data. Apply spectral transformation if feasible (e.g. MSC on Raman spectra)
- Perform PLS2 on e.g. MIR (X) vs. Raman (Y). Apply all relevant validation methods to document their individual behaviour
- Switch the two spectral data types, Raman (X) vs. MIR (Y), and redo the PLS2
- Visualise the joint vs. unique information (Validation Variance) for a selected model complexity (no. of PLS components)

## Campus-scale brewery at AAU EIT (sponsored by Carlsberg)

### Technical summary

2 fermentors approx. 150 L each

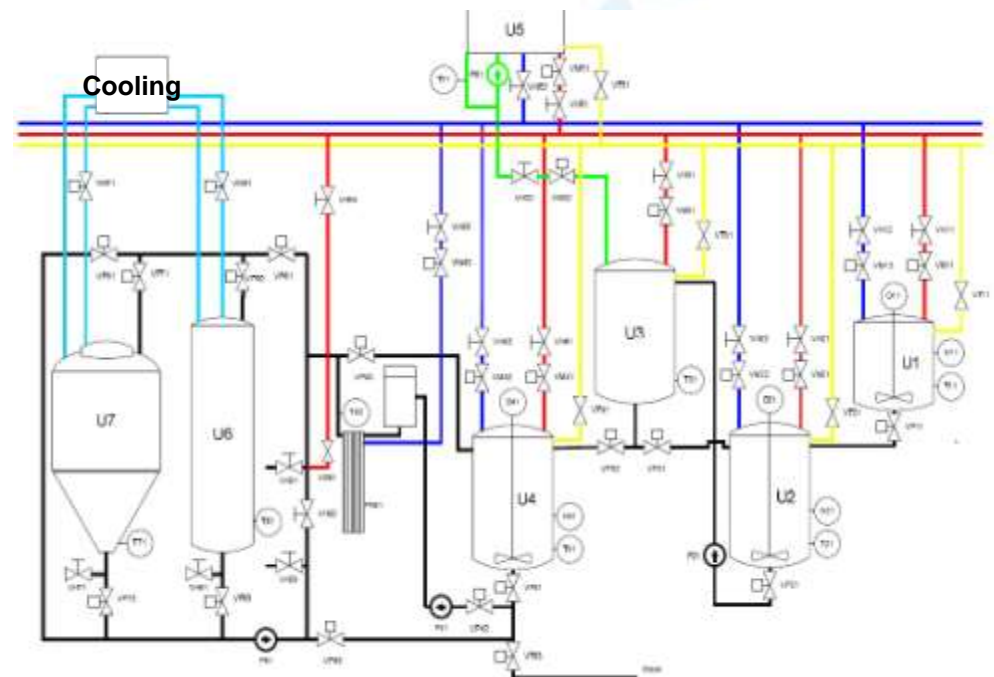
Pre-treatment vessels

Active cooling

Active heating

Ultra-filtration unit

Heat exchanger



Reference: Thygesen and Toft, *Control of brewery*, B.Sc. thesis, AAU EIT, 2006, p. 11

## Plant monitored and controlled through LabVIEW™

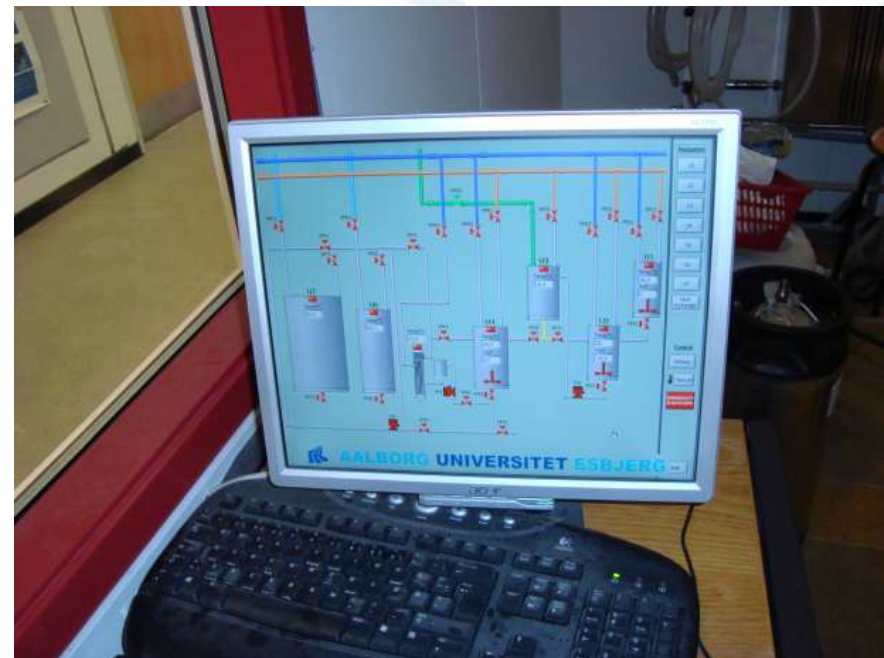
### Future options / ideas

Augment program with chemometric tools:

- online score plots
- early warnings (stuck fermentations)
- real time prediction
- automatic process sampling

### (Hopefully) leading to:

- comprehensive process documentation
- better tasting campus beer...



## Short introduction to the main metabolism

### The production path

Crushing/grinding of barley malt

Wort (sugar solution)

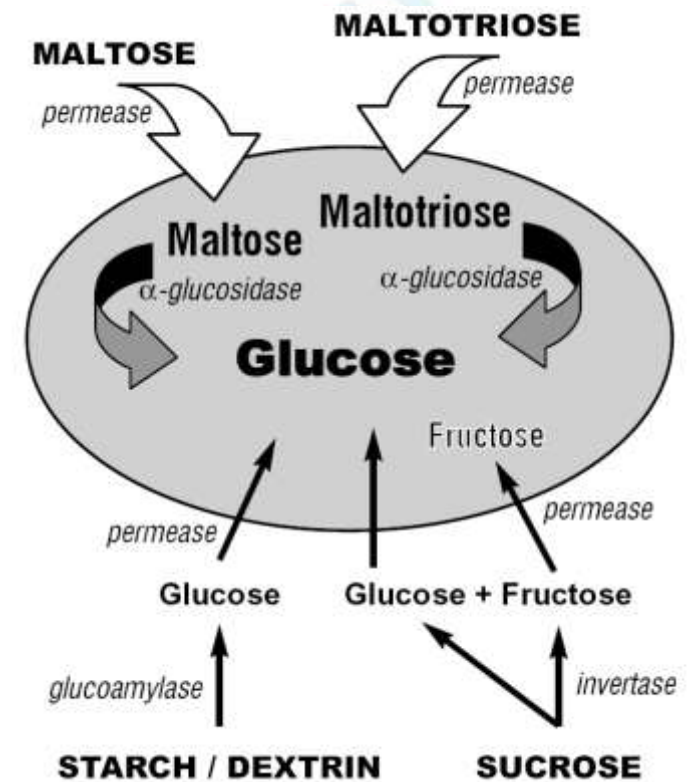
Filtration

Inoculation

Fermentation

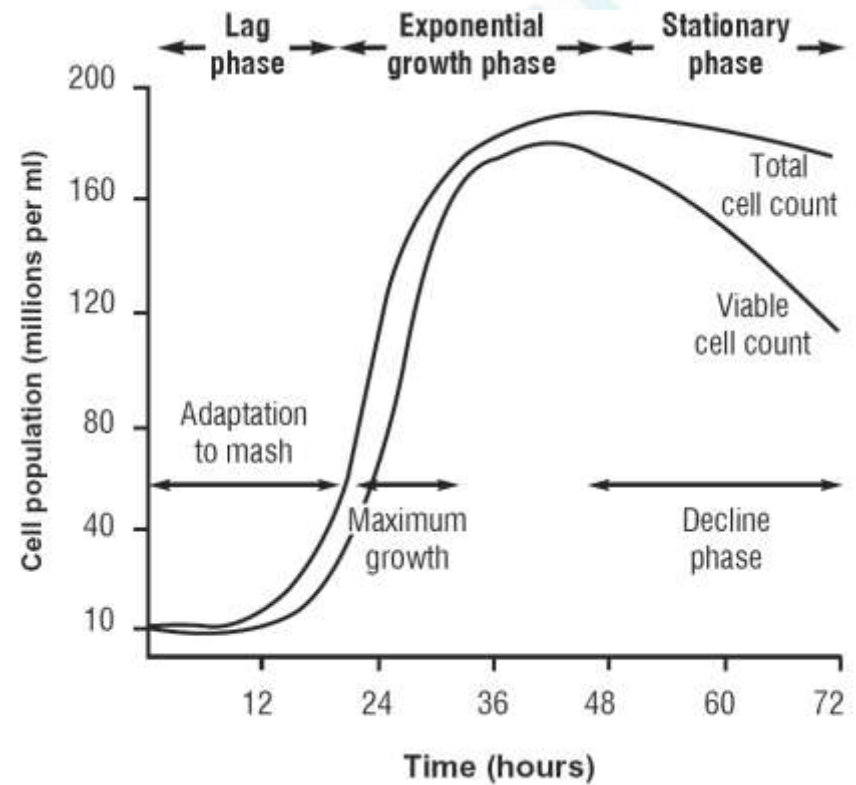
Lager-process

Tapping



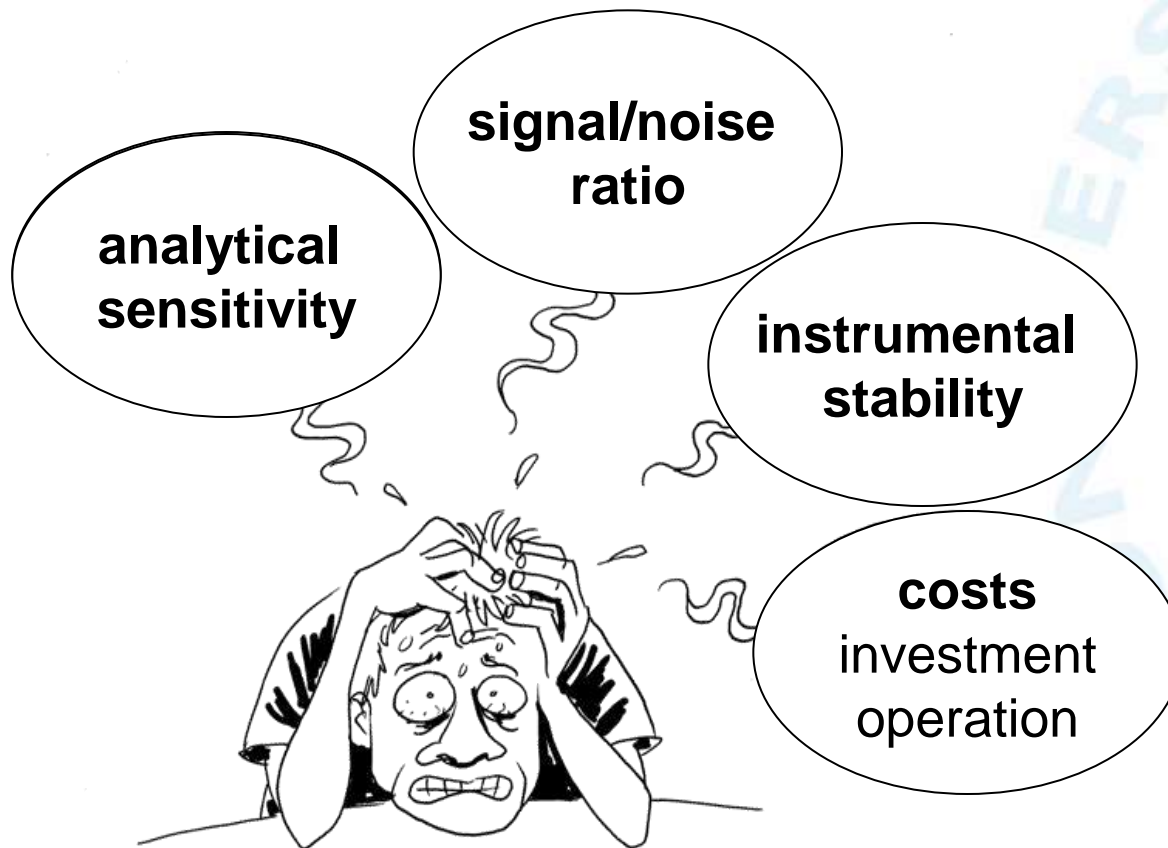
Reference: The Alcohol Textbook, 4th edition (ed. by KA Jacques et al.)

## Typical batch fermentation behaviour



Reference: The Alcohol Textbook, 4th edition (ed. by KA Jacques et al.)

**Spectroscopic implementation:**  
*- A few issues worth addressing*





## Key to unlocking information in signals: *vibrational Eigen modes*

Based on Classical Mechanics and Molecular Quantum Theory the presence of signals from specific molecules in various spectral data types can be predicted

Infrared spectroscopy represents molecular dipole-dipole interactions

Near infrared spectroscopy represents molecular vibrations, stretching, and rotations

Raman spectroscopy represents scatter of incident laser light

In other words: the chosen spectroscopic techniques represent three different physical phenomena

## Features of the chosen spectroscopic techniques

### NIR

- Signals have low intensity
- Wide spectral absorption bands
- Very much affected by the presence of water

### MIR

- Signals have high intensity
- Generally narrow absorption bands
- Some influence from presence of water and carbon dioxide

### Raman

- Signals have very low intensity
- Affected by fluorescence (use excitation laser frequency close to NIR)
- Selective
- Less sensitive to presence of water
- Heats optical sampling area (high power laser excitation technique)

## Fourier Transform InfraRed spectroscopy

**Spectral range**  
4 000 – 650  $\text{cm}^{-1}$

**Resolution**  
8  $\text{cm}^{-1}$

**Co-adds**  
64

**Background**  
Ambient air

**Temperature**  
Ambient (22 °C  $\pm$  1 °C)

**Mode**  
Attenuated Total Reflectance



## Fourier Transform Near InfraRed spectroscopy

**Spectral range**  
9 000 – 5 500  $\text{cm}^{-1}$

**Resolution**  
16  $\text{cm}^{-1}$

**Co-adds**  
64

**Background**  
Ambient air

**Temperature**  
Ambient (30 °C  $\pm$  1 °C)

**Mode**  
Transmission



## Raman laser spectroscopy

**Spectral range**  
3 000 – 200  $\text{cm}^{-1}$

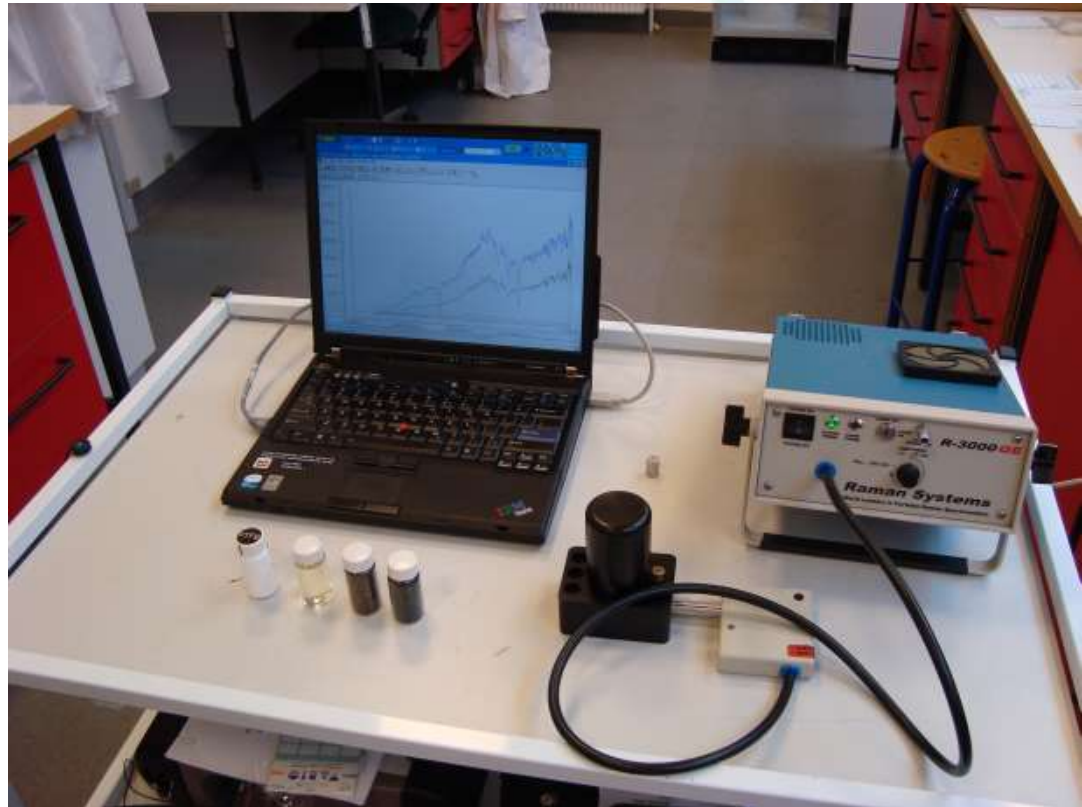
**Resolution**  
10  $\text{cm}^{-1}$

**Integration time**  
1 s

**Laser output power**  
30 mW

**Temperature**  
Ambient (22  $^{\circ}\text{C}$   $\pm$  1  $^{\circ}\text{C}$ )

**Excitation wavelength**  
785 nm



## Wet chemical reference analysis

### Analysed parameters

1. Maltotriose (not metabolised by the yeast applied)
2. Maltose
3. Fructose
4. Glucose
5. Sum of the above
6. Ethanol

### Method

High Pressure Liquid Chromatography with Refractive Index detection, HPLC-RI

### Sample preparation

Dilution with acidic eluent and filtration

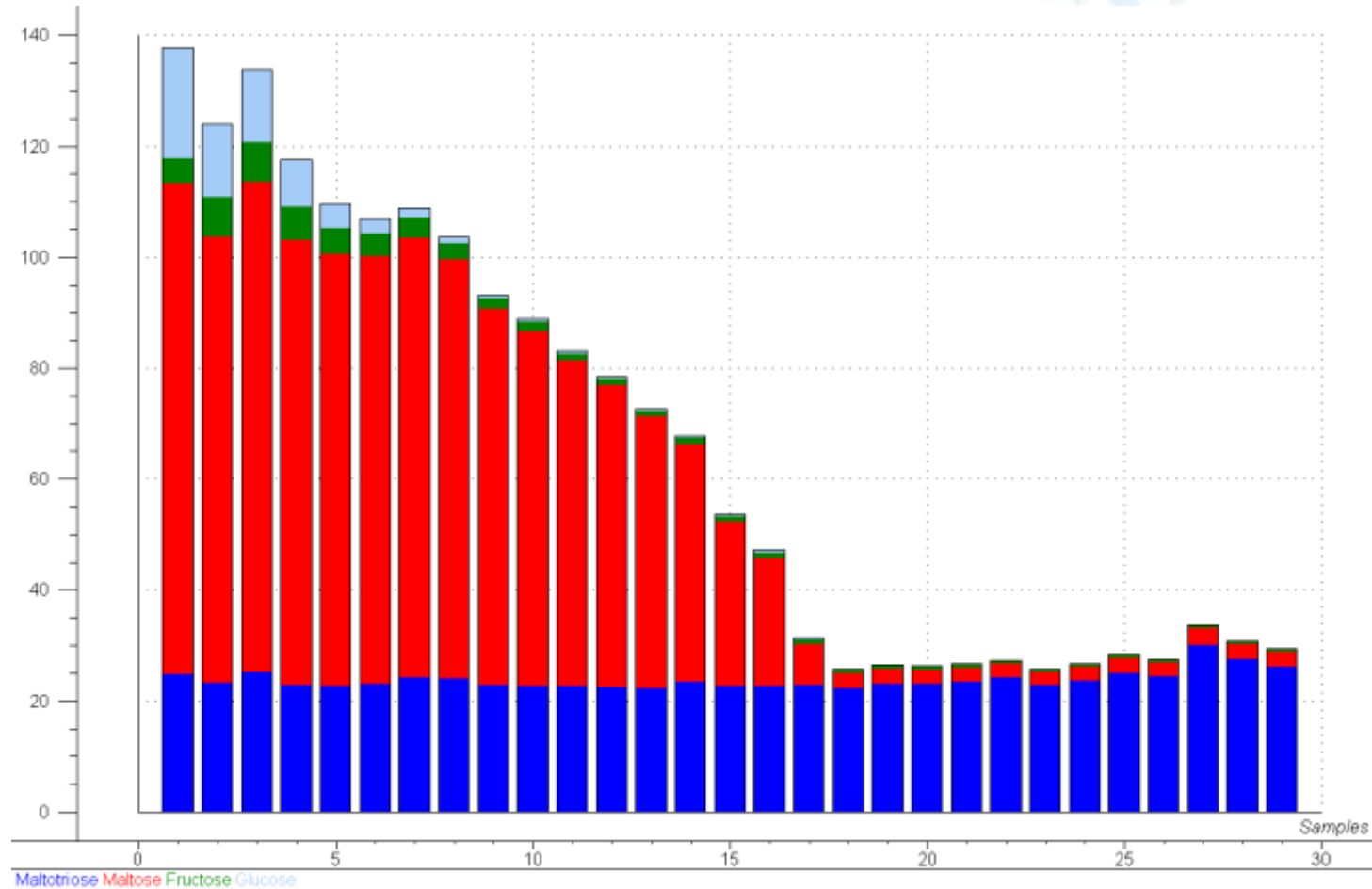
### Analysis time

Roughly one hour per sample (analysed in duplicate)





## Typical sugar utilisation during a fermentation

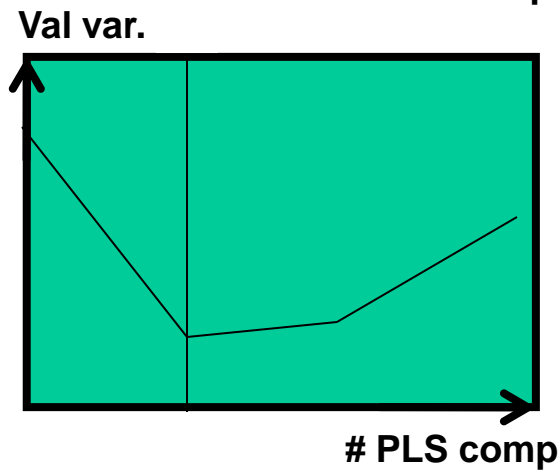
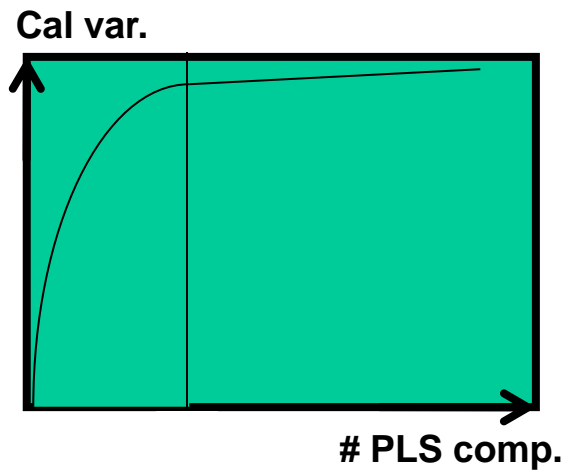


# Sneak preview

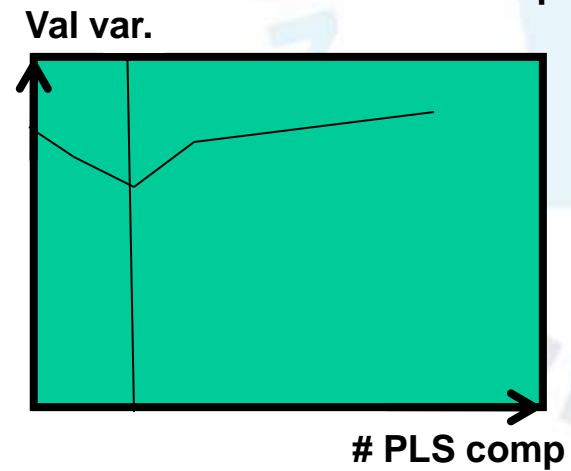
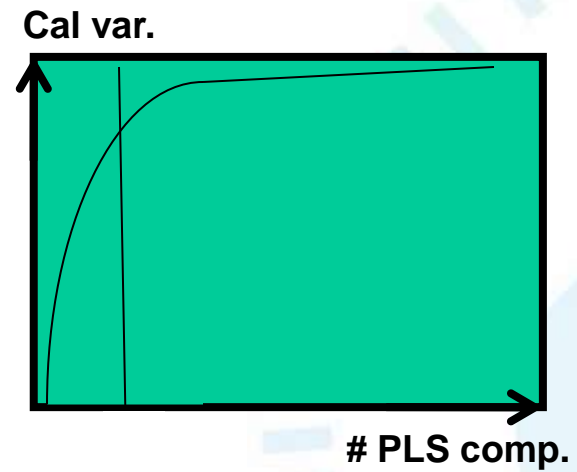
- more in the accompanying paper



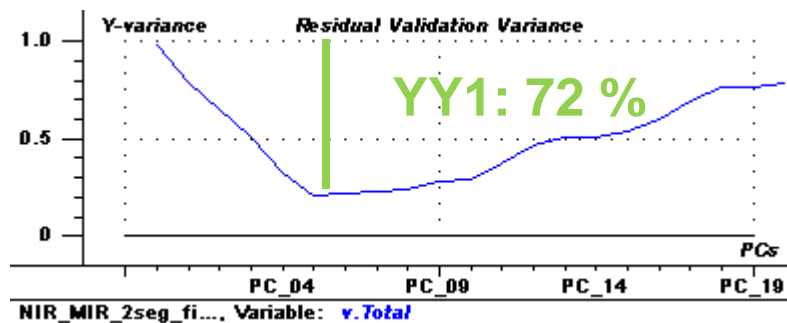
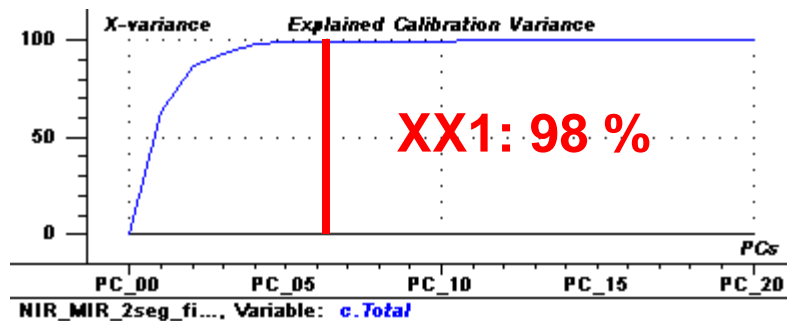
## NIR (X) vs. MIR (Y)



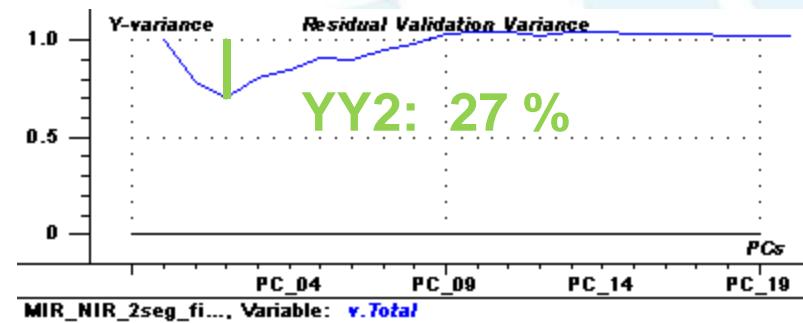
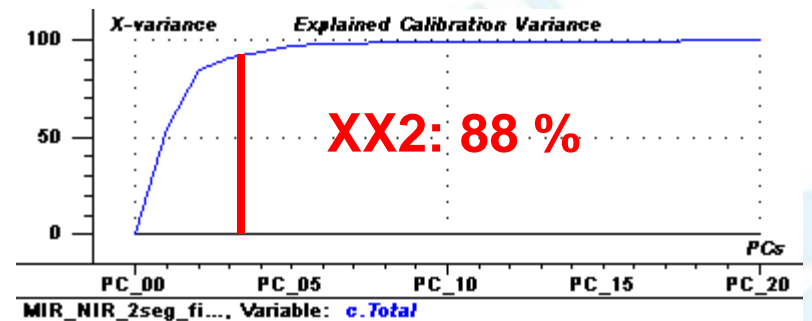
## MIR (X) vs. NIR (Y)



## NIR (X) vs. MIR (Y)



## MIR (X) vs. NIR (Y)



## PLS2 bilateral inter-calibration (A $\rightarrow$ B)

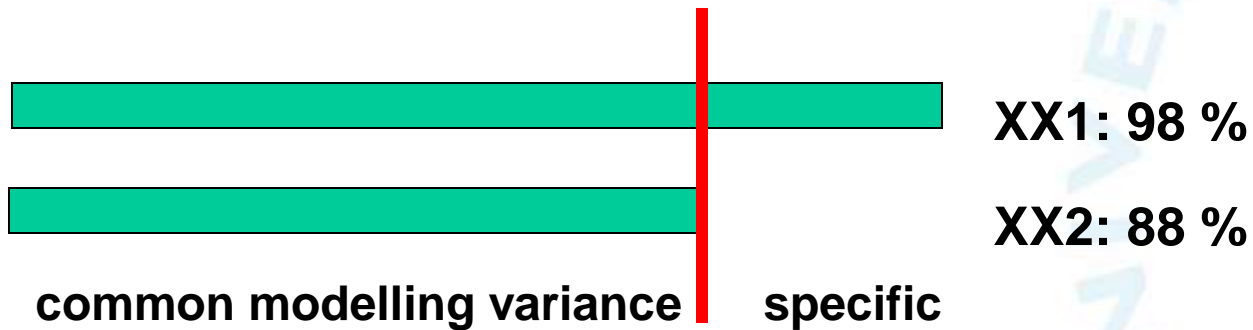
Route diagram:

- 1) Is there a satisfactory model?
- 2) Validation is the answer! (X-val is the right thing **HERE**)
- 3) If the validation is acceptable: cal-variance (X)? : XX1%
- 4) Conclusion: XX1 % explains YY1 %
- 5) Reverse modelling (B $\rightarrow$ A) and repeat procedure
- 6) XX2 % explains YY2 %

## Inter-calibration insight

1: proportion of the common modelling variance (XX %)

2: proportion specific modelling variance



Thank you for your attention

**Avoid spurious correlations  
- practice SAFE validation**



"HEY, I THOUGHT WE WERE WORKING WITH THE SAME DATA..."