

RAPID AND NON-INVASIVE FOOD AND FEED QUALITY MONITORING: THE CHEMOMETRIC POINT OF VIEW

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Raman Spectroscopy (FT-Raman)

Mid Infrared Spectroscopy (FT-IR)

Near Infrared Spectroscopy (NIR)



Instrumentation

Chemometrics

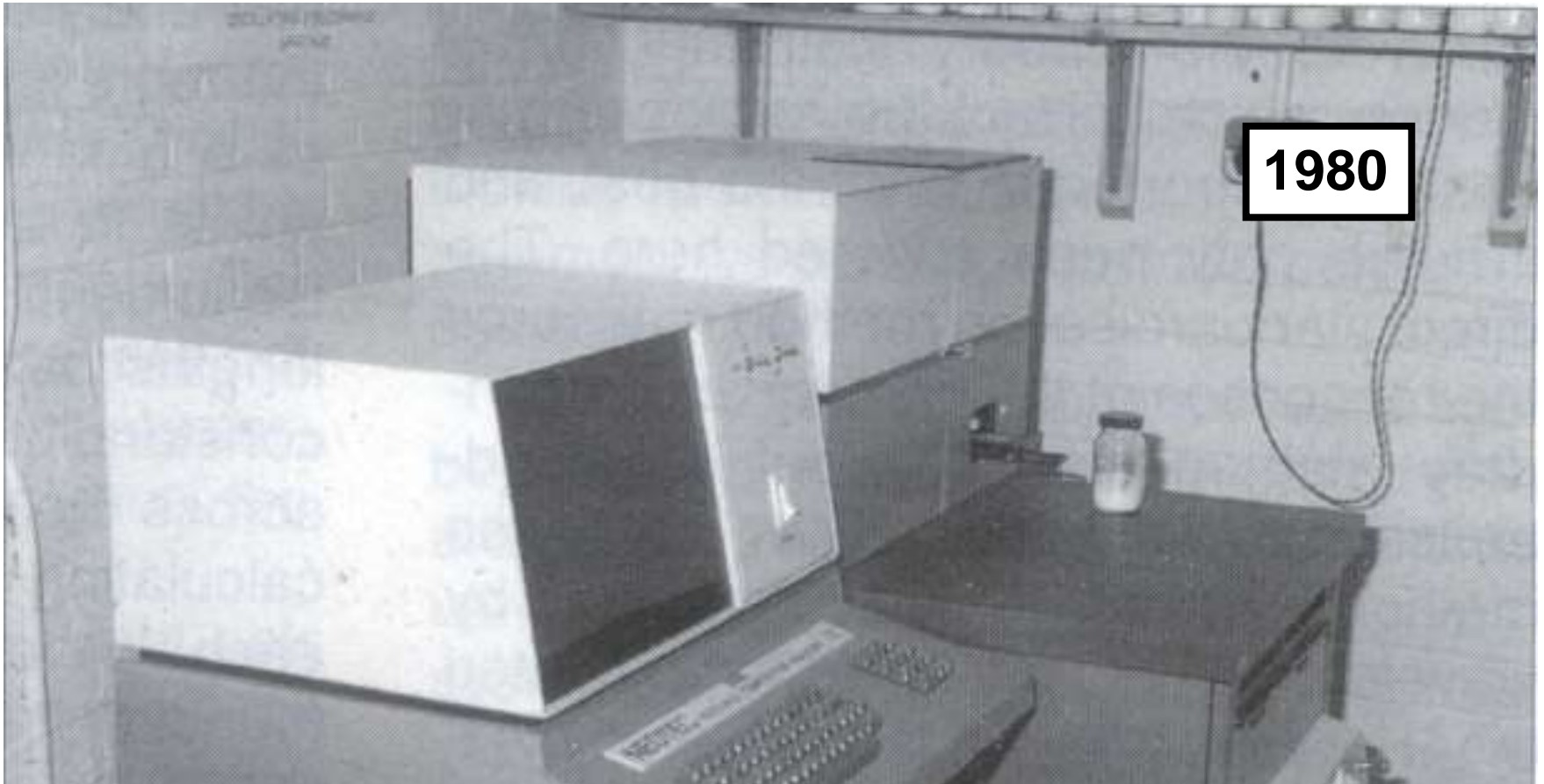
Applications



- use of 'universal' and recognized data bases (models)
(+ wireless & internet)
- chemometrics (new algorithms & computing power)
- increase of on-line installations/measurements
- new low cost portable spectrometers
- new applications for imaging spectroscopy



1980



Underneath, the big box is full of electronics boards and a panel of switches. It is the computer. A data general minicomputer with 32 kb of ram and 2 floppy disk drivers of 8 inches.

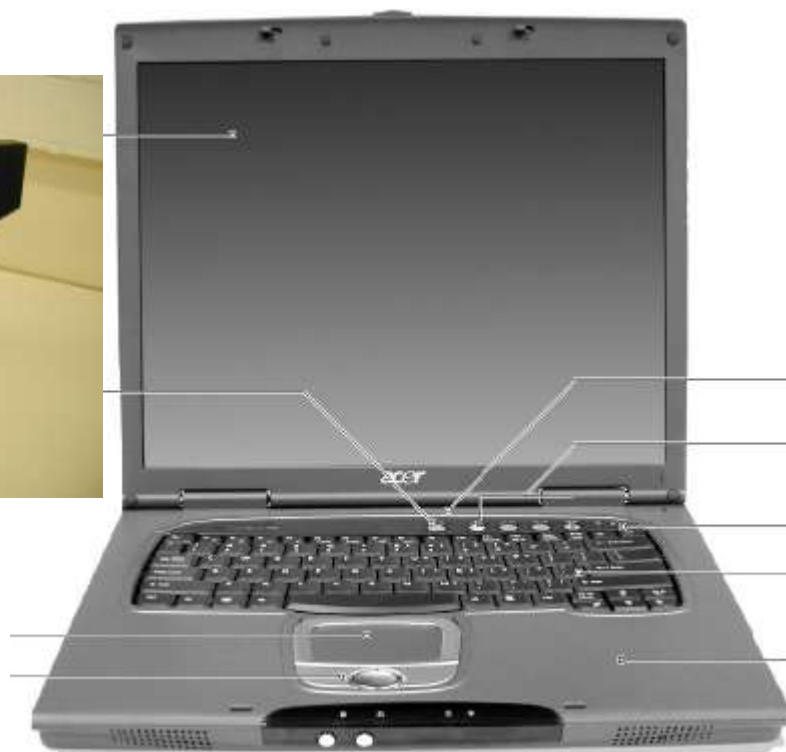
The floppy capacity was 128 kb. One floppy was for the program the second for the data with a maximum of 200 spectra in each.

We calibrated only with a manual step-up regression wavelength by wavelength. If you were lucky, within a day, you had a final model.

P. Dardenne



2007



INSTRUMENT IMPROVEMENTS

- Sources
- Optical components
- Detectors
- Electronics (communication)

Sample presentation:

- large cups
- slurries
- liquids
- on-line (belts, pipes, ...)
 - fiber optics
 - remote scanning



INSTRUMENTATION TRENDS

- miniaturization
→ portable
- imaging
- NIRS instruments to arrive factory-calibrated
- Or user can purchase factory calibrations
based on large sample sets
- NIRS instruments with no calibrations



Low cost spectrometer – farm level



INSTRUMENTATION TRENDS - MINIATURIZATION



Courtesy of Dr John Schenk ■■■

INSTRUMENTATION TRENDS - MINIATURIZATION



Figure # 1. NIR Hand-held Analyzer is taking readings for Brix, pH, TA in vineyard without removing it from vine

**Brimrose Corporation
USA**

INSTRUMENTATION TRENDS - PHAZIR



INSTRUMENTATION TRENDS

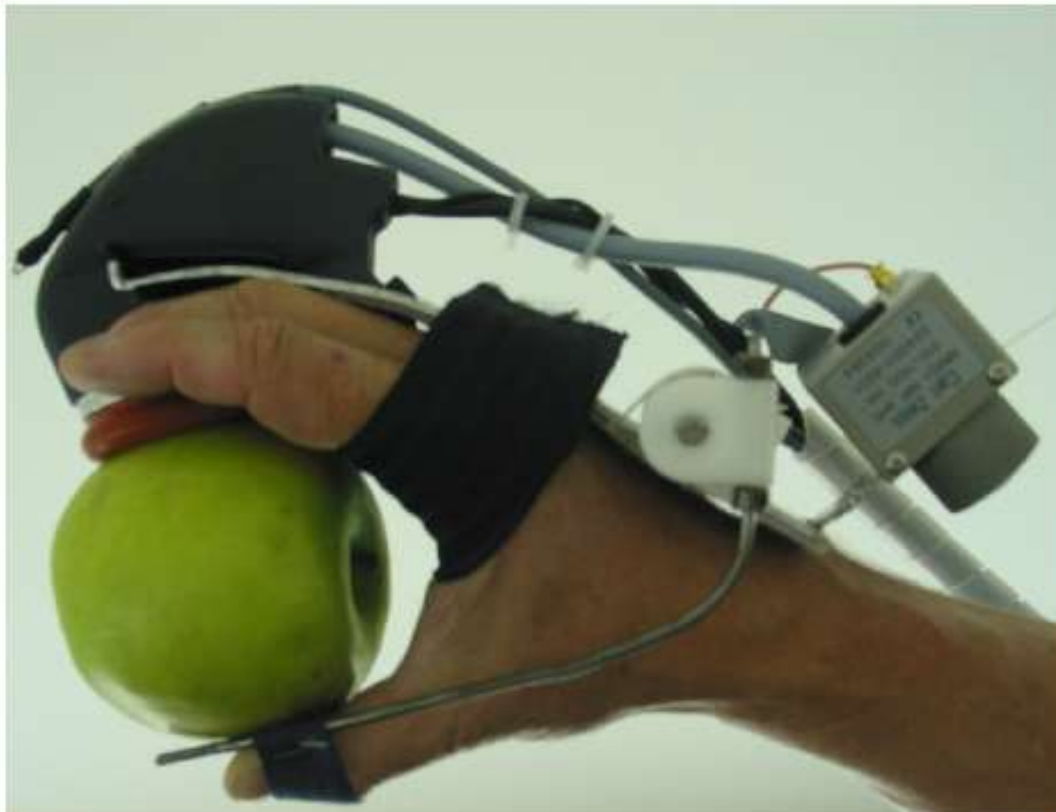


**Gant instrumenté pour mesurer
la qualité des fruits**

GLOVE

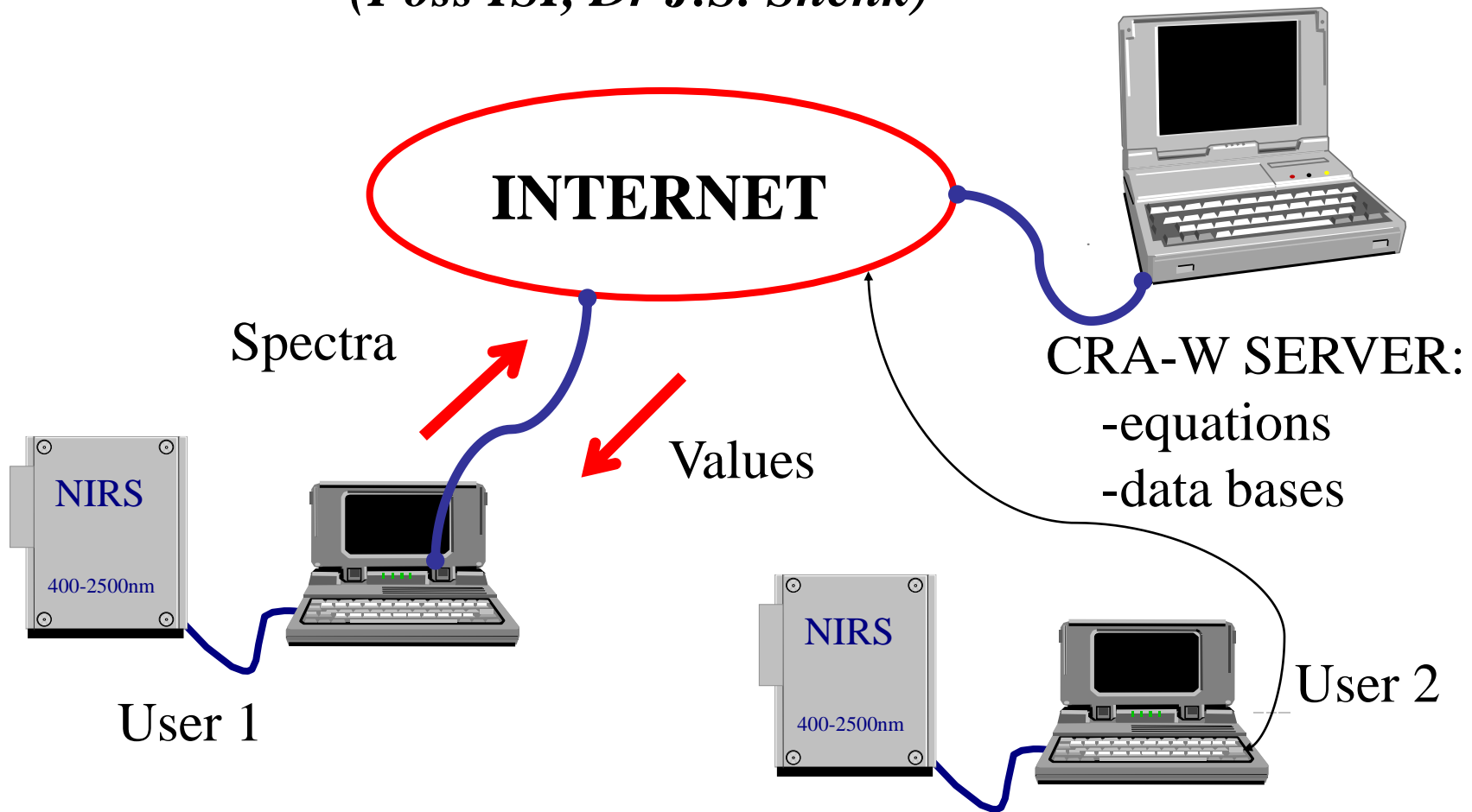
Projet Européen PL 97- 3399 (1998 - 2001)

Partenaires : (Belgique) ; APOFRUIT (Italie); Institut für Agrartechnik Bornim e. V. (Allemagne) VERHAERT (Belgique);
Katholieke Universiteit Leuven - sous la coordination du CEMAGREF - UR Giquel



NETWORKS OF INSTRUMENTS

RINA[®] : *Remote Instrument Near Analysis*
(Foss ISI, Dr J.S. Shenk)



IMAGING



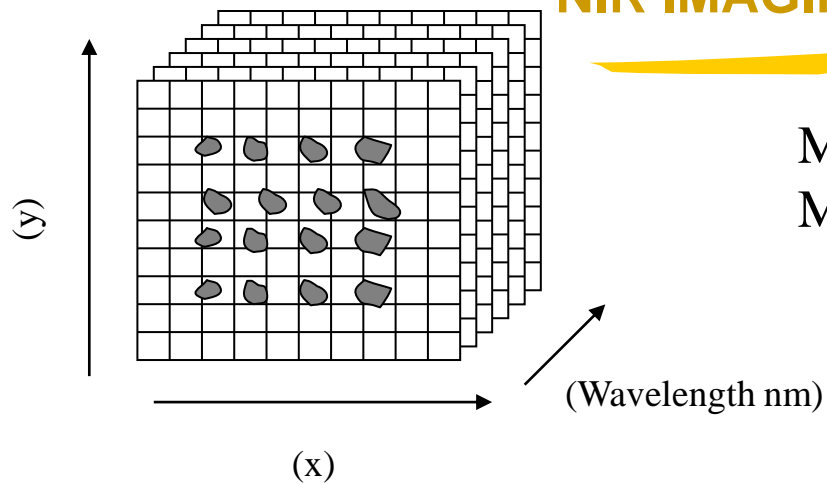
Dedicated session at NIR 2007
Umeå, Sweden
organized by Jim Burger

<http://www.nir2007.com/>



NIR IMAGING SYSTEM

Matrix NIR,
Malvern Instruments Ltd, Malvern, UK



- Camera InGaAs
- 900 - 1700 / 10 nm
- 240 x 320 pixels
- Each pixel: 70 μm *70 μm
- Analyzed surface : 5 cm²
- 76 800 spectra 24 MB
- Analysis time : +/- 10 min



Instrumentation

Chemometrics

Applications



Big gap between published algorithms and what the manufacturers can propose

Quantification

MLR
PLS
ANN
LS-SVM

Variable selection

UVE-PLS
IVE-PLS
AVS-PLS

Classification

PLS-DA
ANN
SVM
SIMCA
k-NN

Outlier detection

Leverage
Residuals
Convex hull
Robust techniques

...

... 

Rapid & specific methods

Local

SVM ...

Methods able to deal with:

large databases

Uncertainty determination

Noise reduction ...



Dealing with...large data bases and rapid methods

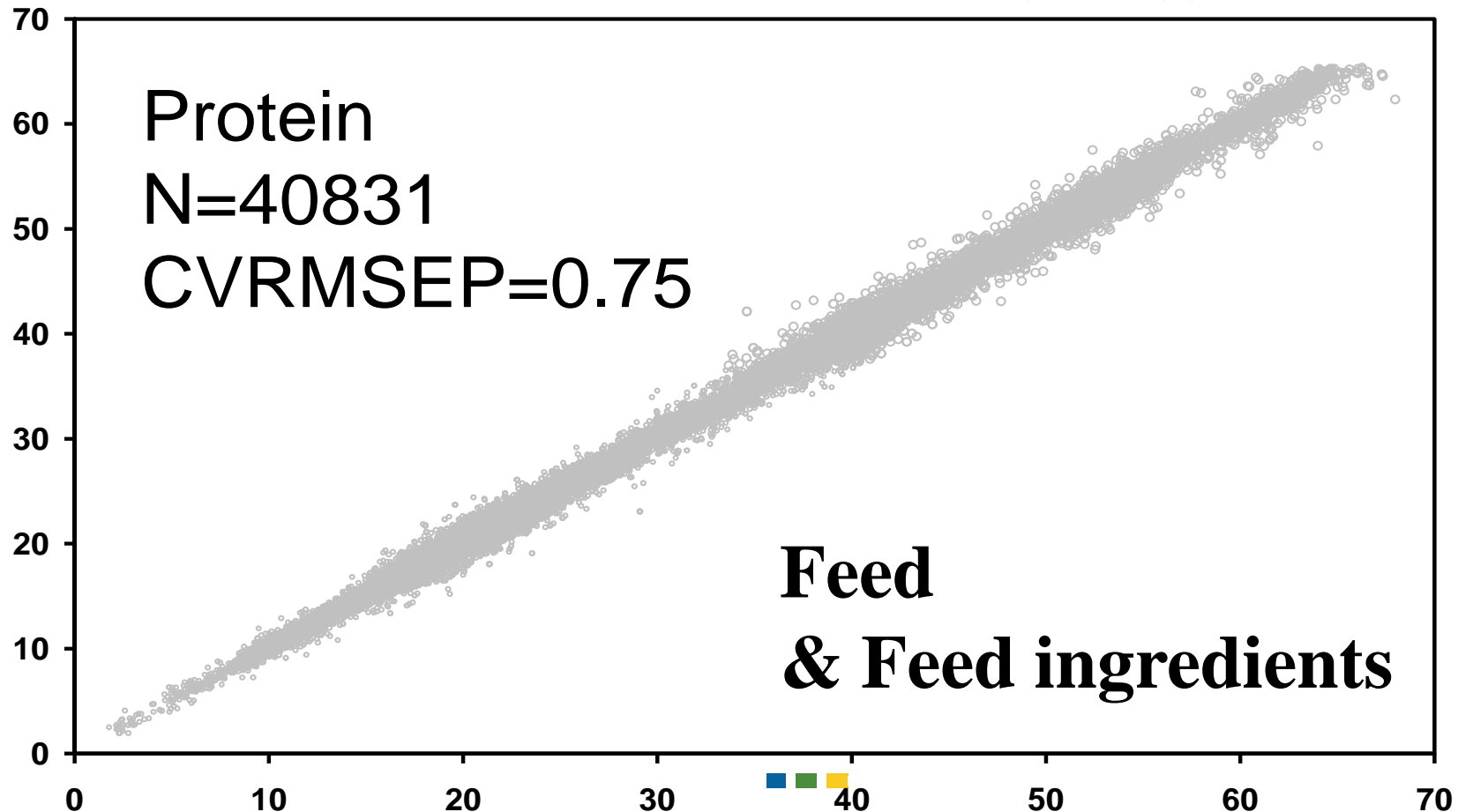
Merging data base: CRA-W & Central Laboratories, UK

Example : **PROTEIN**

N=40831 by LOCAL WinISI III®

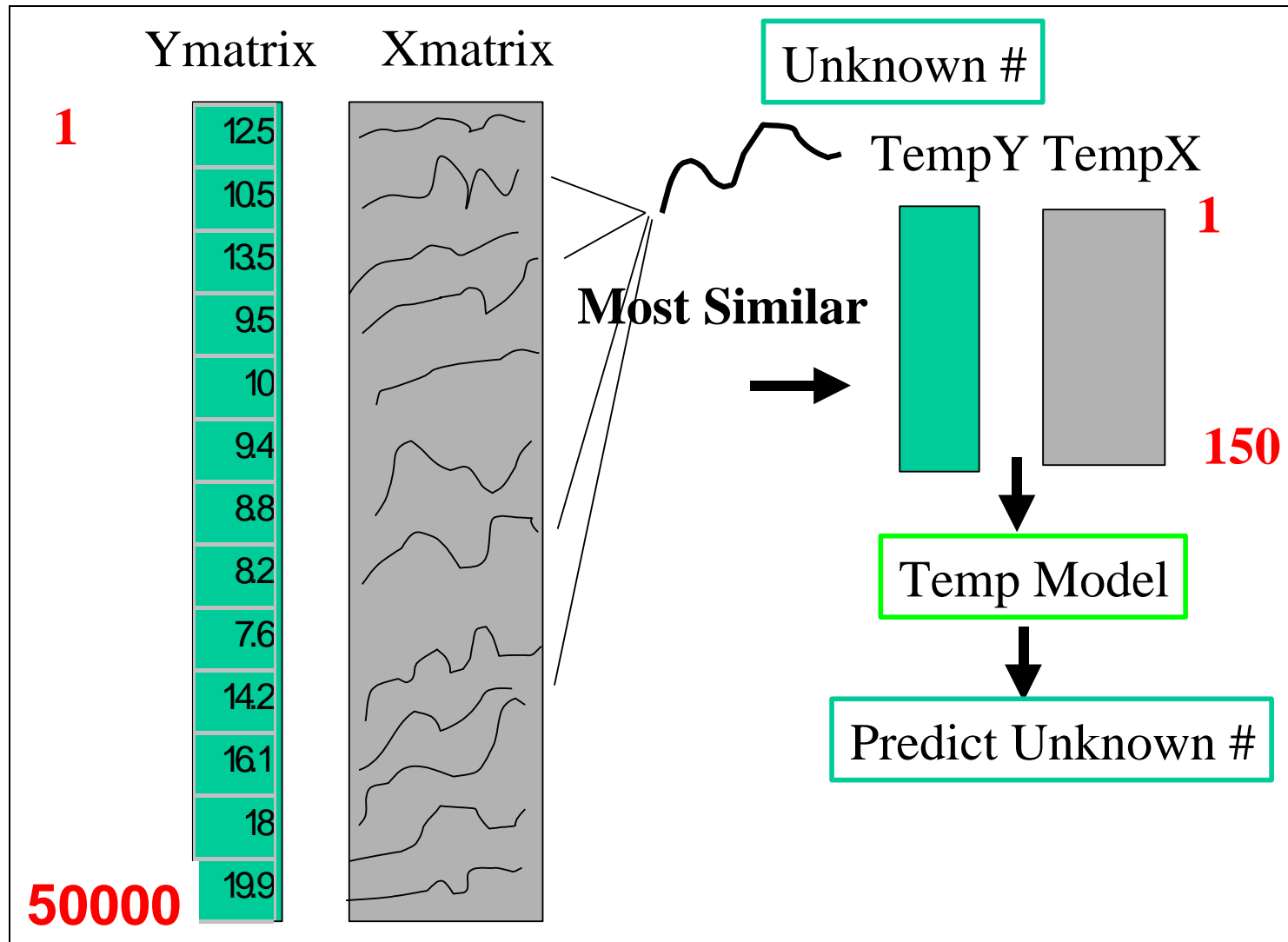
Neighbours samples = 250, Factors ignored=3,

Max factors=33, SNVD 1,4,4 1100-2498/12



Dealing with...large data bases and rapid methods

LOCAL



Dealing with...large data bases and rapid methods

LOCAL

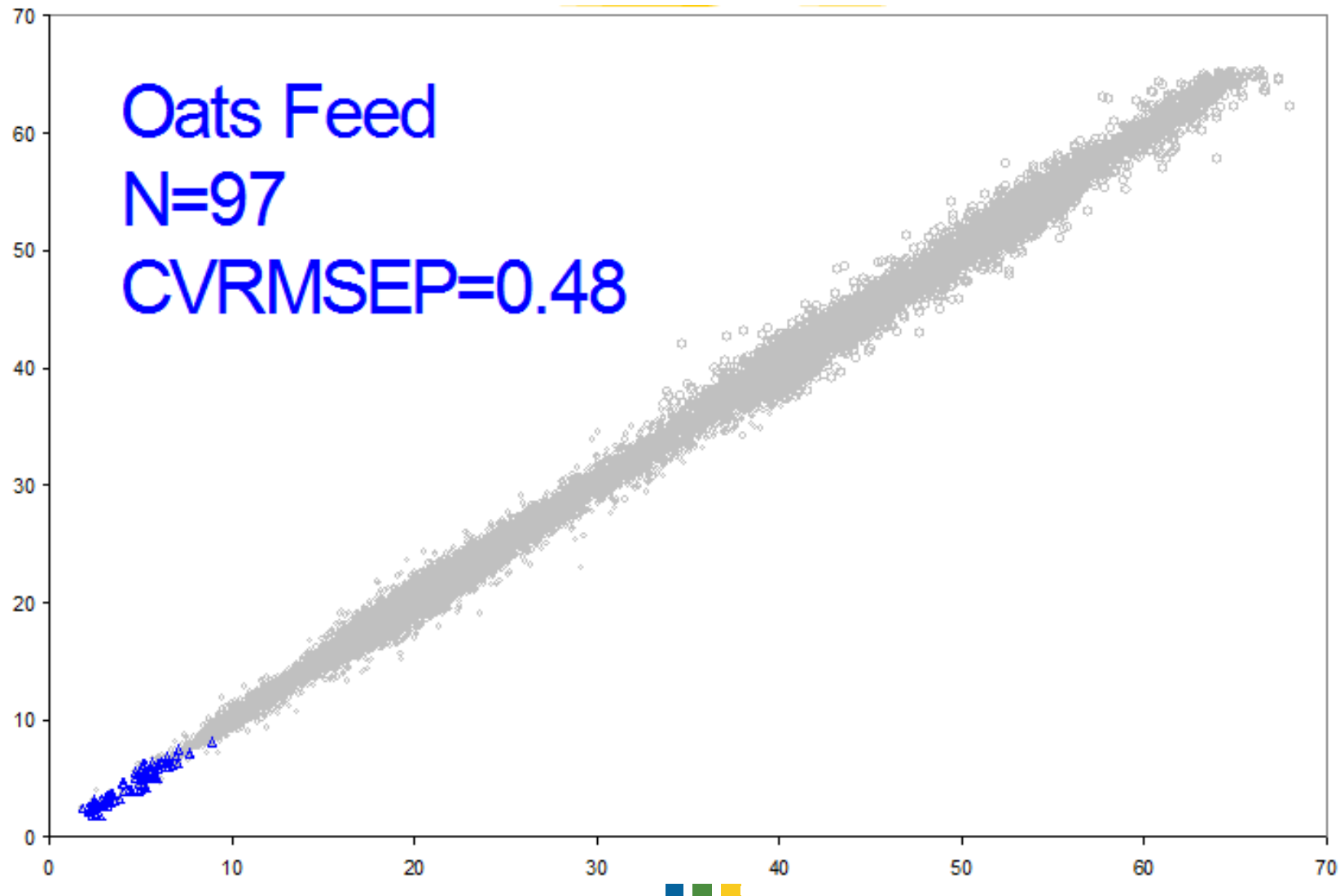
3 parameters: number (N) of closest samples, the maximum number of PLS factors (Fmax) and the minimum number of PLS factors (Fmin).

The final predicted result is a weighted sum of the predicted values from all the models between Fmin and Fmax, values which are weighted according to the standard deviation of the Bcoefficients and to the size of the Xresiduals.

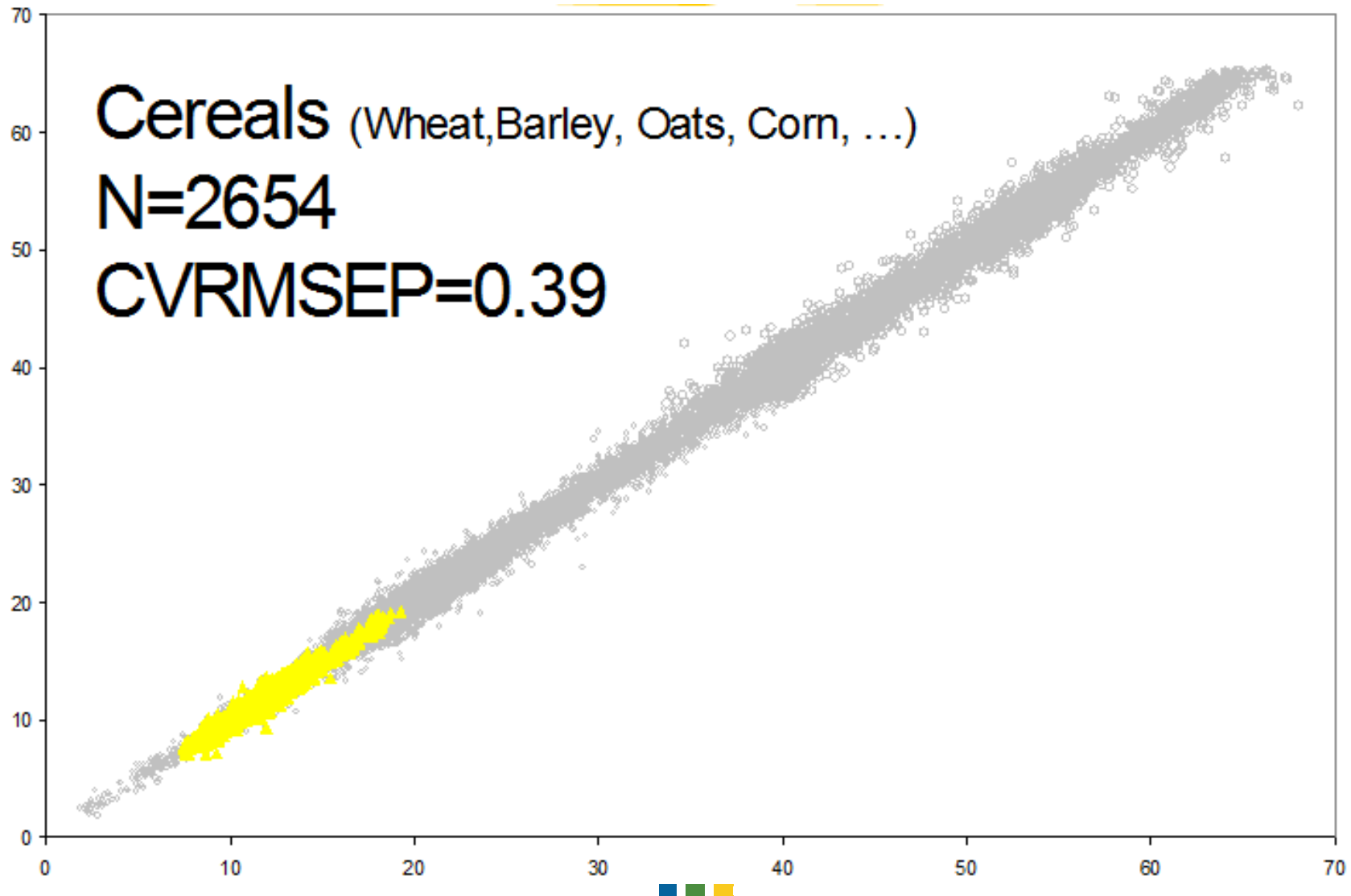
This method is the only one (that we know) which takes information of the unknown sample (the spectrum itself with the use of the Xresiduals) to weigh the predicted values and so to improve the accuracy.



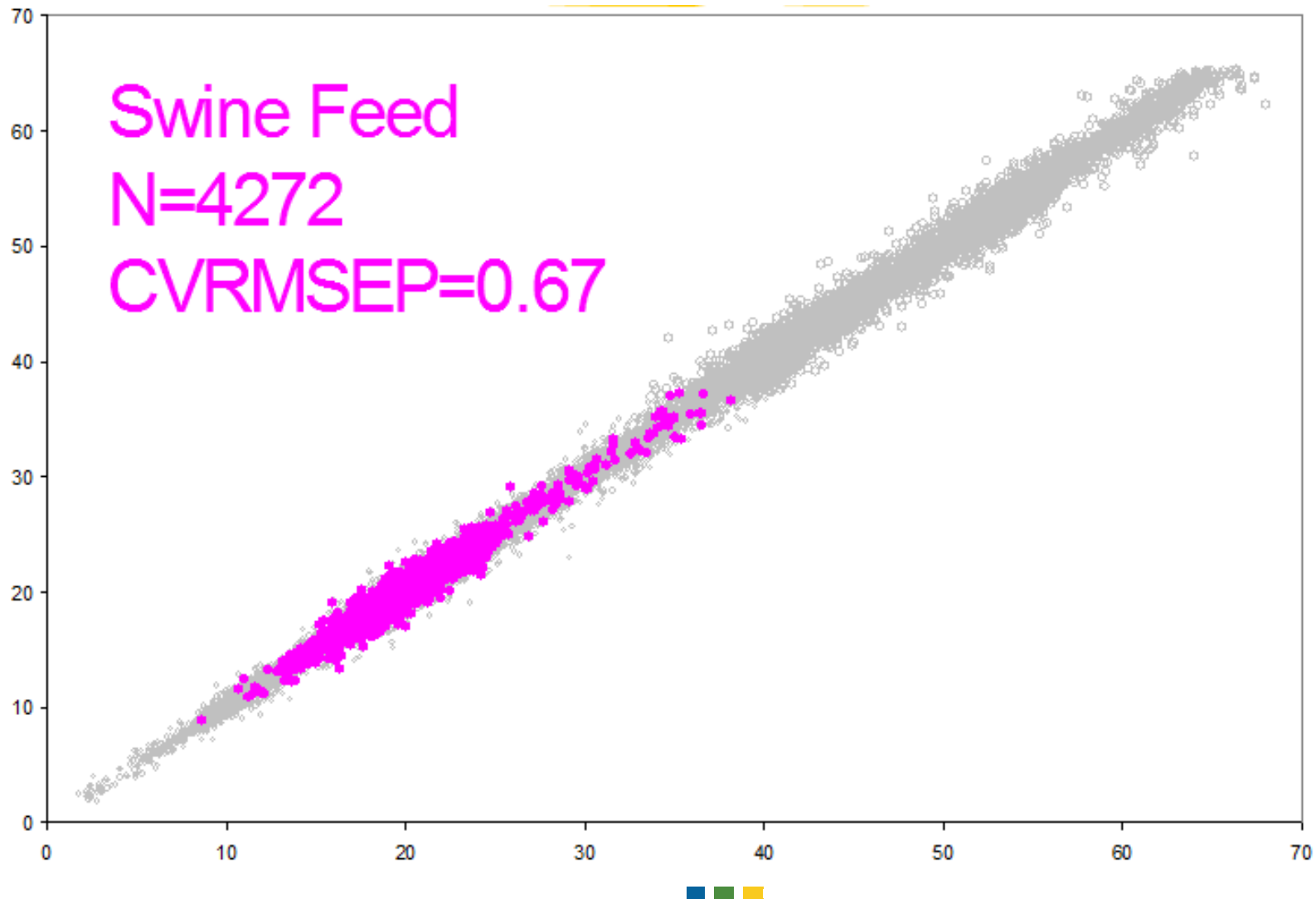
Dealing with...large data bases and rapid methods



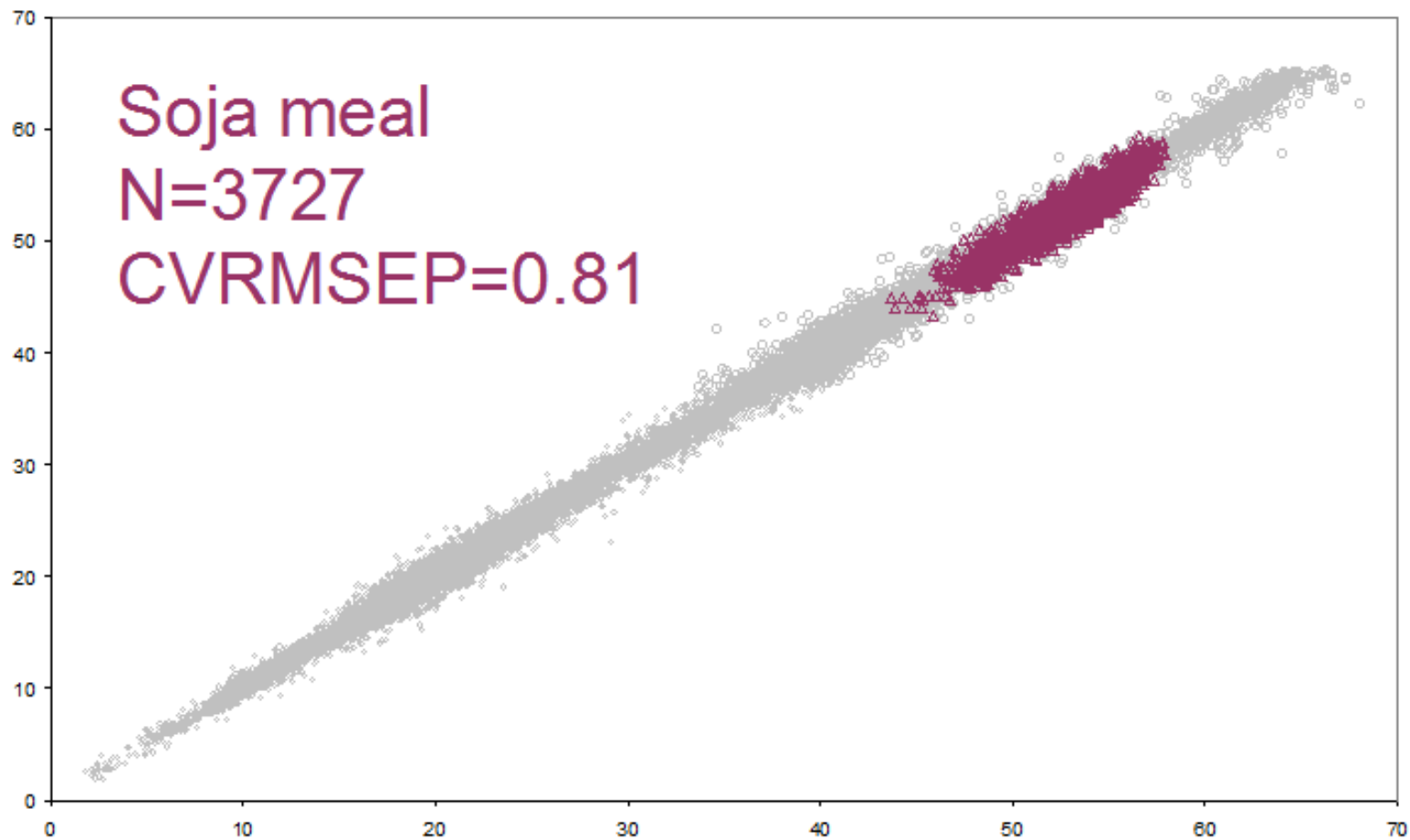
Dealing with...large data bases and rapid methods



Dealing with...large data bases and rapid methods



Dealing with...large data bases and rapid methods



Dealing with...uncertainty

The **uncertainty** of a calculated value is statistically defined as the interval around that value such that any repetition of the calculation will produce a new result that lies within this interval with a given probability

Klaas Faber

‘Estimating Uncertainty in Multivariate Calibration’



Dealing with...uncertainty

$$\text{RMSEP} = \sqrt{N^{-1} \sum_{n=1}^N (\hat{y}_n - y_{n,\text{ref}})^2}$$

\hat{y}_n = prediction for sample n

$y_{n,\text{ref}}$ = associated reference value



Dealing with...uncertainty

The result (RMSEP) is a constant measure for prediction uncertainty that cannot lead to prediction intervals with correct coverage probabilities (say 95%).

A crucial assumption is that the reference values are sufficiently precise; this is certainly not always true (octane rating, classical Kjeldahl) - often the prediction is even better than the reference value.

High intrinsic variability of RMSEP estimate requires N to be large.



Lecture of Ornella Preisner, IBB: Bootstrapping...

Sample-specific prediction uncertainty of the NIR analyses

$$s(\hat{y}_i - y_i) = \left[(1 + h_i) \cdot SEC^2 - S_{ref}^2 \right]^{1/2}$$

Faber and Bro, *Chemom., Intell. Lab. Syst.* 61, 133 (2002)

Fernandez Pierna & al., *Chemom., Intell. Lab. Syst.* 65,281 (2003)

$$SEP_{actual}^2 = SEP_{observed}^2 - SEL_{ref}^2$$



Dealing with...noise

Regression model $y = Xb + e$

$$b = (X^T X)^{-1} X^T y$$

A good estimate for ***b*** is required

It must provide:

- A good fit to y
- good predictions for unknown samples



Dealing with...noise

The accuracy of b (and therefore the accuracy of \hat{y}) estimated by NIR models depends on the quality of the reference method.

The performance of the reference laboratory methods limits the reliability of the NIR calibrations.

A crucial assumption in multivariate calibration is that the reference values are sufficiently precise. This is not always true!



Dealing with...noise

L. Munck, Univ. Copenhagen

“Data is ‘the king’ that rules chemometric modelling”

Phil Williams, PDK Grain, Canada

“Just as the spectra are the heart of NIRS technology, the reference test is the heart of NIR applications”



Dealing with...noise

We aim to prove that one can still have good calibrations with quite poor reference values or with a large quantity of noise added.

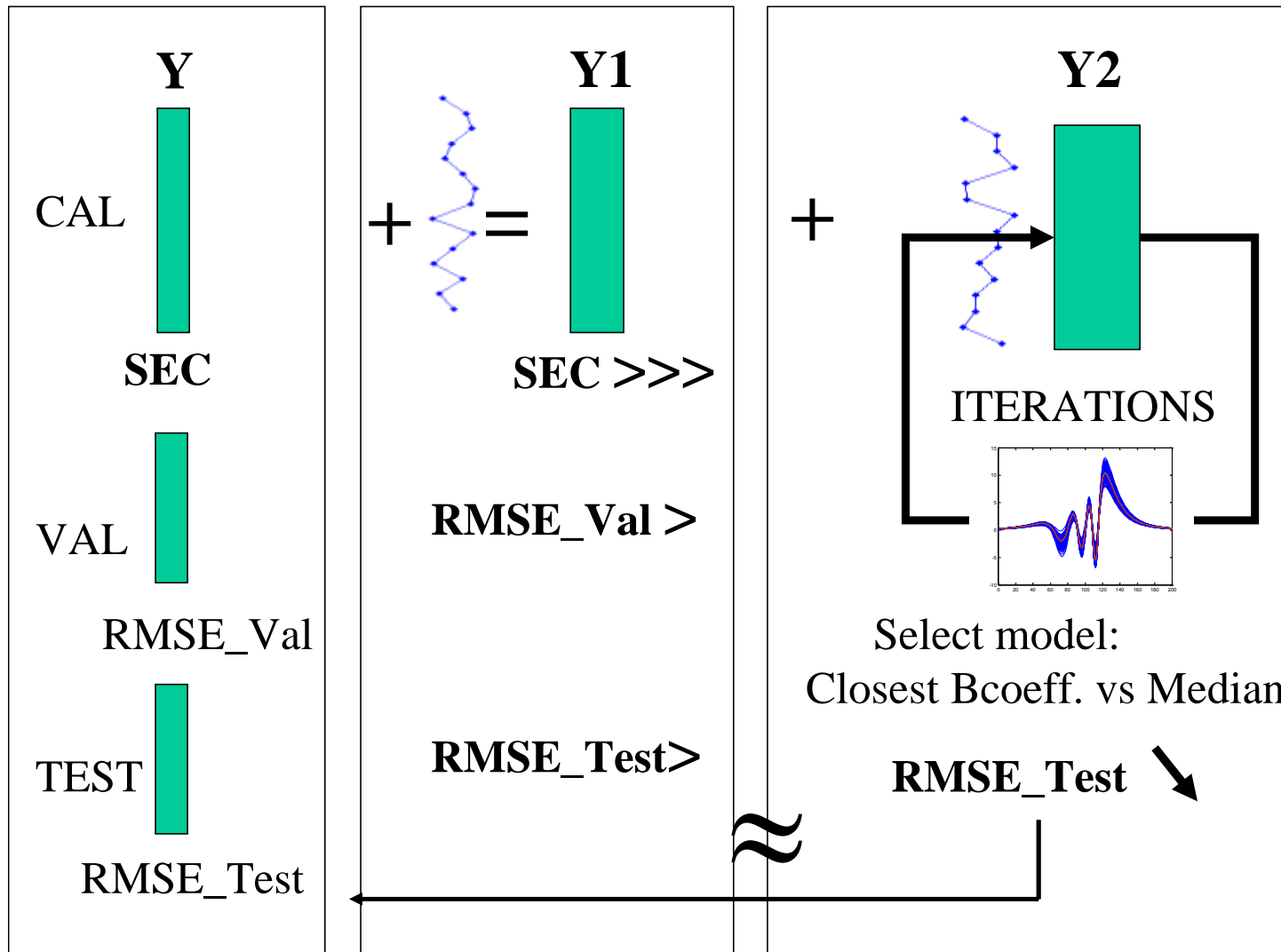


NAPLS

Dardenne P. & Fernández Pierna J. A. 'A new method to improve the accuracy of the NIRS models: NAPLS: Noise addition PLS method'
J. Near Infrared Spectrosc. 14, 349-355 (2006)



NAPLS



NAPLS – FORAGE DATA SET

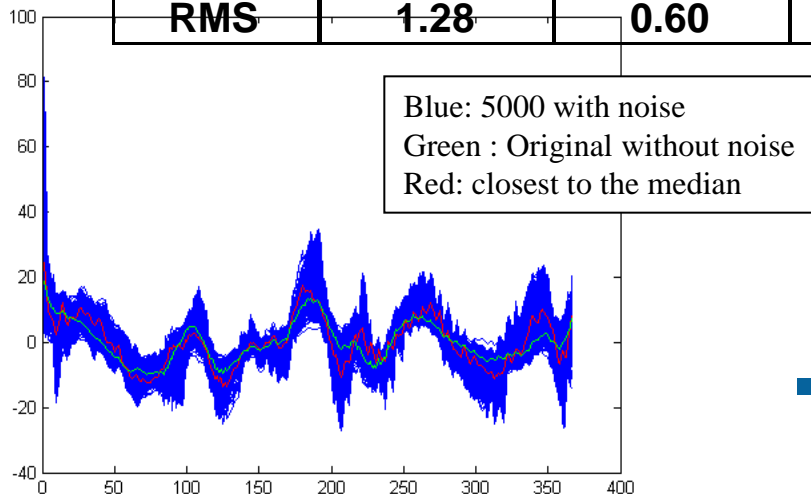
Run	RMScal_noise	RMSval	RMStest	RMStest_final	% difference test
1	1.65	0.77	0.86	0.59	31.61
2	1.52	0.68	0.82	0.67	17.39
3	1.41	0.71	0.80	0.62	22.36
4	1.62	0.68	0.86	0.62	28.12
5	1.67	0.91	1.02	0.80	21.41
6	1.49	1.05	0.97	0.68	30.02
7	1.64	0.80	0.80	0.58	26.88
8	1.50	0.77	0.75	0.55	27.47
9	1.68	0.76	0.77	0.60	22.43
10	1.67	0.67	0.64	0.56	12.30
RMS	1.59	0.79	0.84	0.63	24.43



RMStest_orig=0.60

NAPLS – WHEAT DATA SET

Run	n=100 RMScal_n oise	n=100 RMSval_n oise	n=2450 RMStest_n oise	RMStest_fi nal	% difference test
1	1.34	0.50	0.47	0.39	15.51
2	1.28	0.57	0.50	0.42	16.43
3	1.25	0.63	0.53	0.36	32.70
4	1.24	0.69	0.59	0.42	28.71
5	1.27	0.64	0.55	0.41	26.22
6	1.27	0.65	0.55	0.37	33.67
7	1.36	0.59	0.53	0.36	30.94
8	1.23	0.62	0.55	0.37	32.04
9	1.21	0.53	0.47	0.38	19.64
10	1.31	0.57	0.50	0.40	20.15
RMS	1.28	0.60	0.53	0.39	26.05



RMStest_orig=0.50

Near Infrared Spectroscopy (NIRS)

Instrumentation

Chemometrics

Applications



- Fertilizers
- Soils
- Seeds & Phyto-sanitary Protection
- Crop monitoring (N)
- Precision Agriculture
- Nutritive value (feed & forages)
- Technology (flour, baking quality,...)
- Authentication (olive oil, meat, honey,...)
- Fruits
- Transformed Products (meat, dairy, juices,...)
- Bio-fermentation monitoring



•NUTRITIVE VALUE OF FEED

CHEMICAL COMPOSITION & DIGESTIBILITY

- **Moisture – DM**
- **Ashes – OM**
 - + P, Ca, K, Mg
- **Fat**
 - + FA profile
- **Proteins (N)**
 - + AA profile
- **Fibres**
 - (cellulose, NDF, ADF, ADL)
- **Starch**
 - + amylose - amylopectin
- **Total Sugar**
 - + sugar profile
- **OMD**
 - in vivo, in vitro, enzymatic*

Feed Ingredients

Cereals & by-products
Wheat bran
Soyameal
Sugarbeet pulp
Animal protein (MBM)
.....

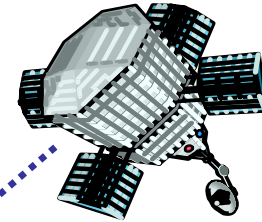
Complete feed

Cattle
Swine
Poultry
Pet food

PRECISION AGRICULTURE

**NIR
Spectrometer**

GPS



Parameters:

DM

....

Protein

Fibre

OMD

•

•



CONCLUSION

- use of 'universal' and recognized data bases (models)
(+ wireless & internet)
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EVENTS

Belgian Chemometrics Society



**Workshop on
Multivariate Image
Processing**

**November 8, 2007
Gembloux - Belgium**



**9th Belgian Chemometrics
Symposium and
presentation of the D.L.
Massart Award in
Chemometrics 2008**

April 11, 2008, Brussels

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**FEED SAFETY
International Conference 2007
Methods and Challenges
27th and 28th November 2007**

Centre de Congrès du Beffroi
Namur - Belgium



<http://safedpap.feedsafety.org/fs2007/>