



Università degli Studi di Padova

Forage Calibration transfer from laboratory to portable instruments: challenges and applications

Paolo Berzaghi and Xueping Yang

University of Padua GraiNit srl





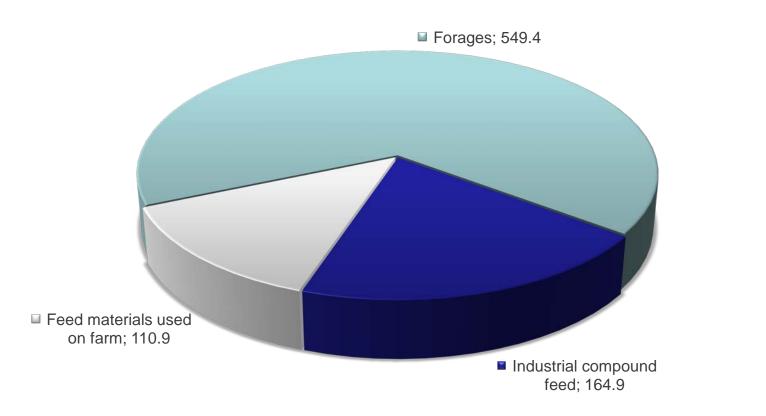
- There is a problem (Lab bench⇒Handhelds)
- I'm not satisfy of results
- Any Idea?





Europe Feed and Forage market

Livestock sourcing in feed in the EU+UK (825 mt. in 2020)



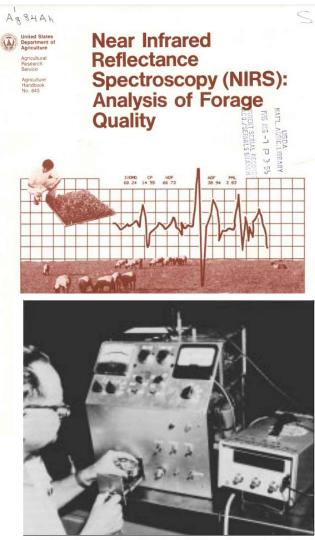
Source: FEFAC, DG AGRI

mt = million tonnes



When it all (NIR speaking) Started

- Composition (protein, fat, moisture, starch, Energy,)
- Biological properties (Digestibility, palatability)
- Since the early time, NIR has established itself as the analytical method of choice for farms, feed company, feed and forage trading



Karl Norris operating the first NIR analyser for oil and soya beans (1968).



Laboratory NIR instruments





Traditional Lab Challenges

- Too slow!!
 - Sampling
 - Taking the sample to a lab
 - Preparing the sample (drying & grounding)
 - Preparing the sample in a scanning cup
 - Selecting the appropriate calibration
 - ➤NIR Scanning and predicting
 - ≻Getting results back to the user



At farm applications



- Need to take the NIR to the samples
 - To have rapid answers
 - Implement immediate corrections when needed
- Need to have the technology more accessible
 - Lower cost





Portables/handhelds

Take the instrument to the product to analyze

Sampling

 Taking the sample to a lab
 Preparing the sample (drying grounding)
 Preparing the sample in a scanning cup
 Selecting the appropriate calibration
 NIR Scanning and predicting
 Getting results back to the user



Portable instruments

DietSensor



Le scanner de poche SCIO fonctionne avec l'application DietSensor compatible aux smartphones IOS et Android

(SCIO wit developper par Consumer Physics)







Fairy tales....??

- Lellsper



Calibration transfer -Standardization

- There are different ways to transfer calibration (Fearn 2001):
 - Creating a robust calibration 'insensitive' of the differences between instruments
 - 2. Adjust the results (Slope and Bias) using the same prediction model
 - 3. Adjust the spectra (standardization)



1 Laboratory vs 3 Handhelds

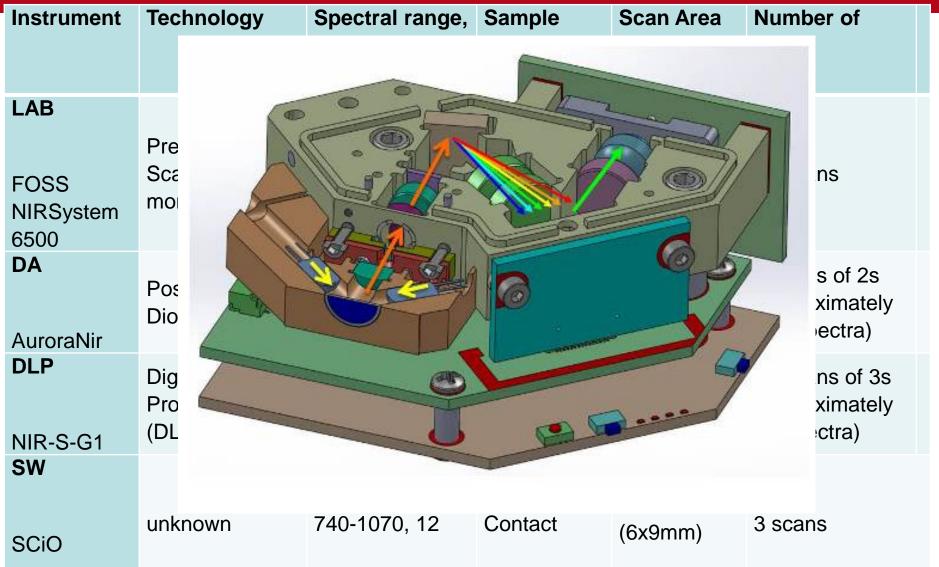


LAB = Foss6500	~50000€
DA = Diode Array	~ 15000€
DLP = Dig. Light Processing	~ 3000€

SW = Short Wave ~ 300€

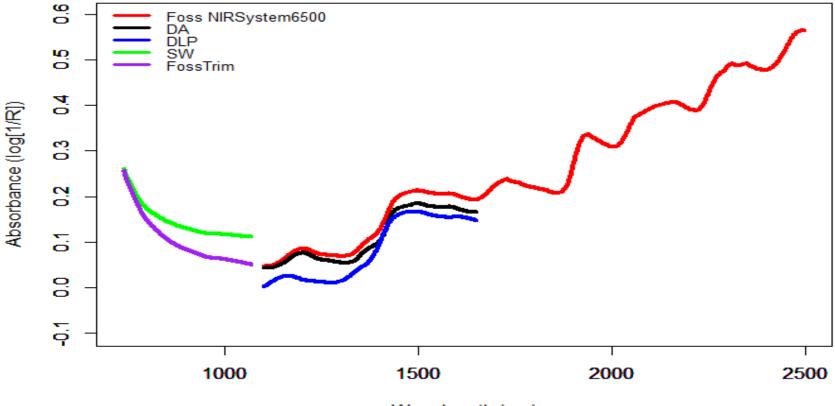


Features of the instruments





Average Spectra



Wavelength (nm)



Forage samples

- Samples were from 8 agronomic trials at Cornell University (Dr. Jerry Cherney)
- Dried (60 °C) and ground (1mm, Wiley

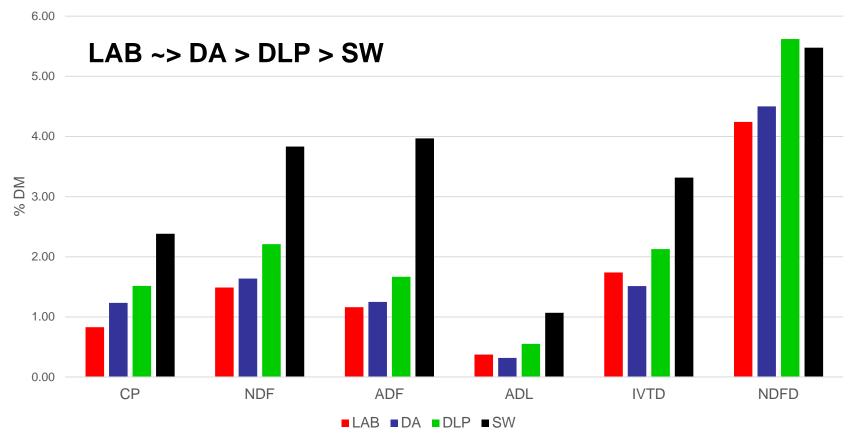
Trial no.	Study	Year		Alfalfa, n.	Grasses, n.
15-45	1	2015	Calibration	100	94
15-85	2	2015	Calibration	40	33
16-67	3	2016	Test-set	88	54
16-69	4	2016	Test-set	63	63
16-91	5	2016	Calibration	83	83
16-92	6	2016	Calibration	52	60
17-68	7	2017	Calibration	60	48
17-93	8	2017	Calibration	126	81
Total				612	516

Berzaghi et al., 2021



Comparing performances

SEPs Alfalfa samples



Berzaghi et al., 2021



Slave

Università degli Studi di Padova



Transfer to handhelds Masters







DietSensor

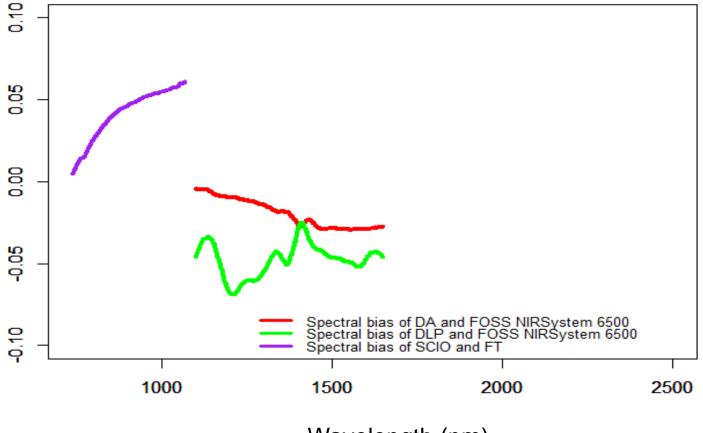


Le scanner de poche SCIO fonctionne avec l'application DietSen: compatible aux smartphones IOS et Andreid

LAB calibration developed over the years



Average difference from LAB spectra

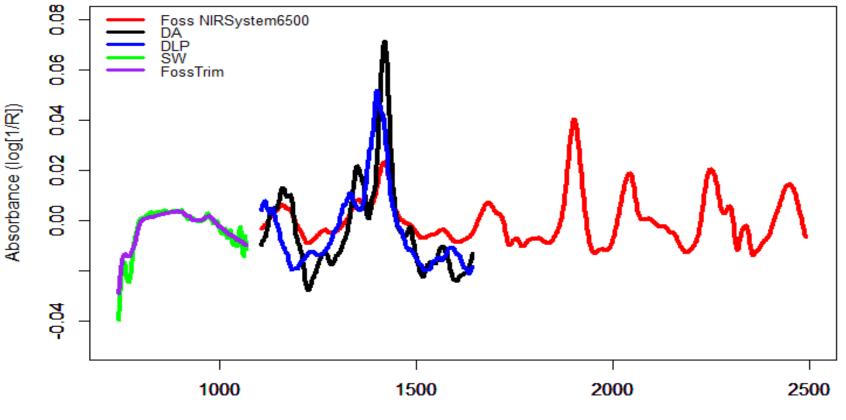


Wavelength (nm)



Spectra after math treatments

SNV, Detrend, SG 1d, 7gap



Wavelength (nm)



Hardware differences

0.8

0.6

0.4

0.2

0

-0.2

-0.4

-0.6

-0.8

-0.6

-0.8

SNV, Detrend, SG 1d, 7gap

1598

1498

1398

1298

1198

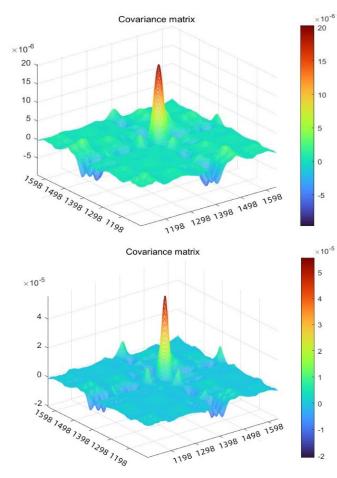
1198

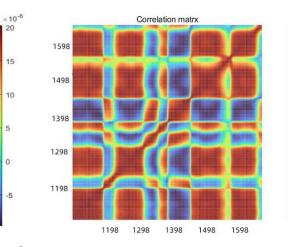
1298

1398

1498

1598





Correlation matrx

LAB (1100-1650 nm)

0.8 0.6 0.4 0.2 0 **DA (1100-1650 nm)** -0.2 -0.4



Hardware differences

0.8

0.6

0.4

0.2

0

-0.2

-0.4

-0.6

-0.8

0.8

0.6

0.4

0.2

0

-0.2

-0.4

-0.6

-0.8

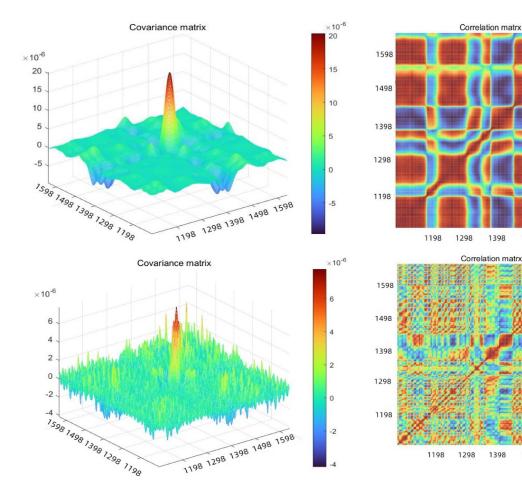
1498

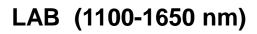
1498

1598

1598

SNV, Detrend, SG 1d, 7gap





DLP (1100-1650 nm)



Hardware differences

0.8

0.6

0.4

0.2

0

-0.2

-0.4

-0.6

-0.8

SNV, Detrend, SG 1d, 7gap

×10⁻⁵

1.5

1

0.5

0

-0.5

-1

×10⁻⁵

5

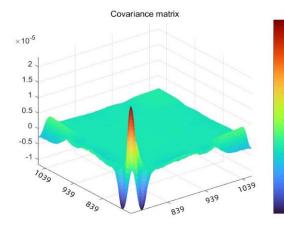
2

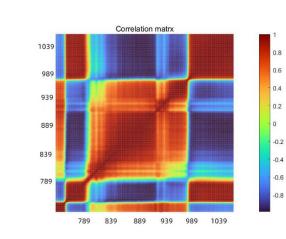
0

-1

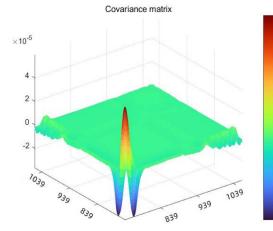
-2

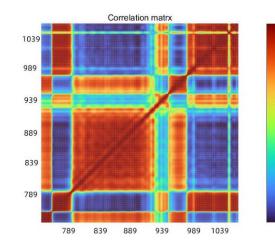
-3





LAB (740-1070 nm)





SW (740-1070 nm)



Calibration transfer

Spectra correction:

- Spectra Bias Correction: simply apply the average spectral difference to the Master instrument
- Piecewise Direct Standardization: 1 Comp, Windows 3 datapoints (RNIR)
- Shenk & Westerhaus: WinISI

Inoculation

 Adding "real spectra and chemistry of Master instrument AFTER Spectra correction



Università decli Studi

DI PADOVA

Spectra correction challenges

- All instruments must scan the same samples, possibly untouched
- Sample presentation may affect spectra
 - Foss samples are placed in ring cup cuvette
 - The same cuvette could not be scanned on the DLP
 - Instruments were at different locations and spectra were collected at different times with possible changes of samples



Using prediction as reference values

 What about using NIR prediction to develop a calibration for a new instrument?
 ③

Isn't it cheating? ☺

UNIVERSITÀ DEGLI STUDI

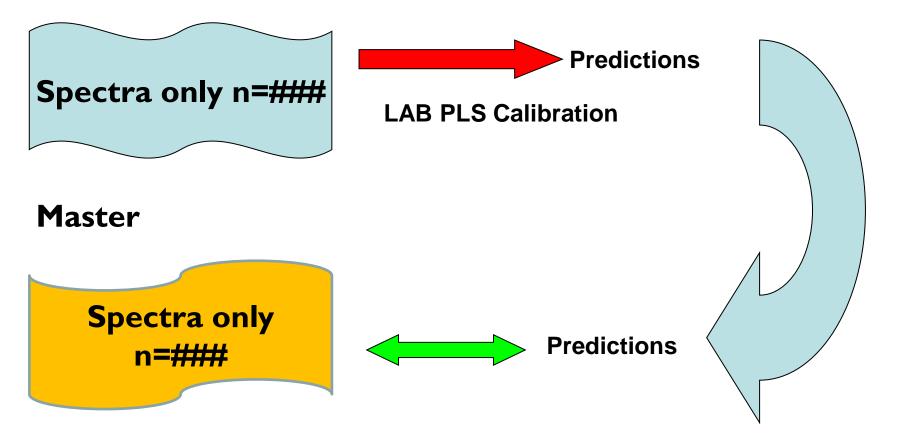
DI PADOVA

• Or is it a legitimate use of prediction data?



Using prediction as reference values

LAB (Slave)





Using prediction as reference values



Master Spectra and 'chemistry' n=#### PLS1 **Constituent**= β_0 + β_1 + β_2+ β_n



Using prediction as reference values

Advantages

Università degli Studi di Padova

- Cheap and quick
- Doesn't involve complicated chemometric
- Can be used for instrument with different wavelength
- Disadvantages
 - Still need to scan the same samples on both Master and Slave
 - It will be easier to overfit new Master PLS models



Forage samples

Samples were from 8 agronomic trials at Cornell University (Dr. Jerry Cherney)

Trial no.	Alfalfa, n.	Grasses, n.
Total	612	516
LAB Calibration	295	285
Standardization	40	33
Test-set	151	117
Inoculation	42	42
LAB NIR predictions	124	114

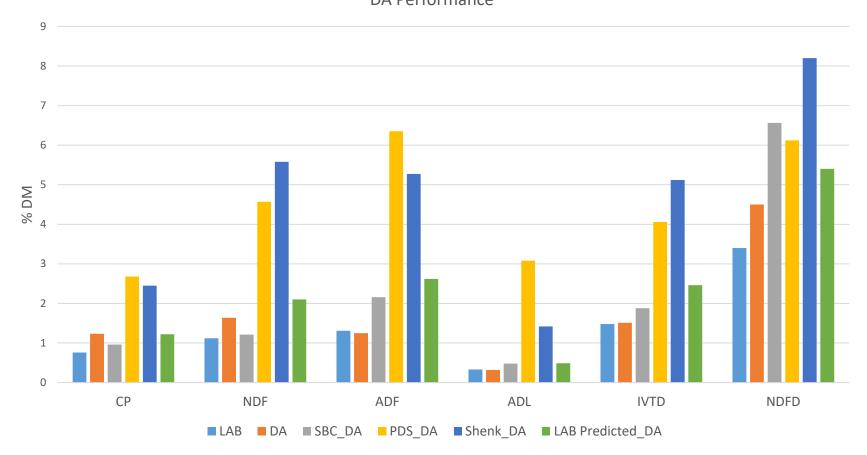
CP = Crude Protein; ADF = Acid Detergent Fiber; NDF = Neutral Detergent Fiber; IVD = In-Vitro True Dry matter Digestibility; NDFD = NDF digestibility





DA SEPs Alfalfa

SBC (+26%) <LAB_pred (+36%) <PDS (+157%) <S&W (+168%) DA Performance

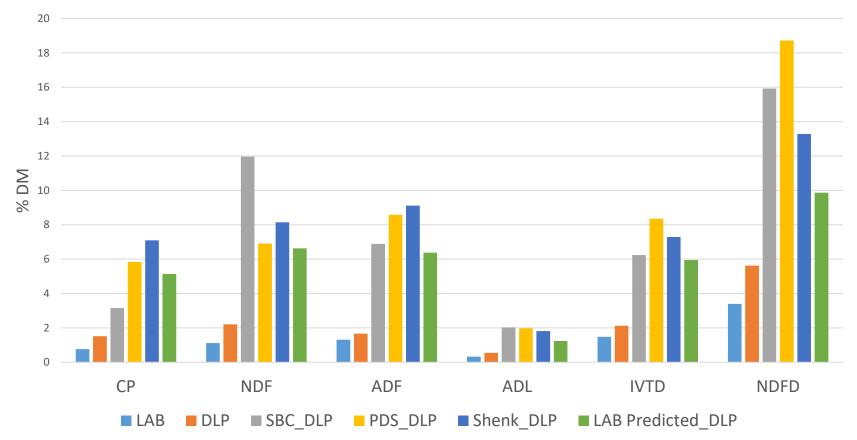




DLP SEPs Alfalfa

LAB_pred (+157%) <SBC (+237%) <S&W (+241%) <PDS (+268%)

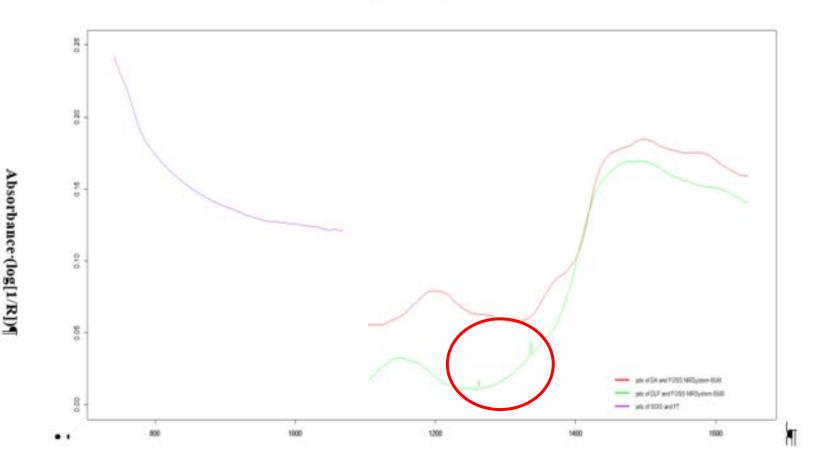
DLP Performance





PDS spectral artifact

PDS spectra

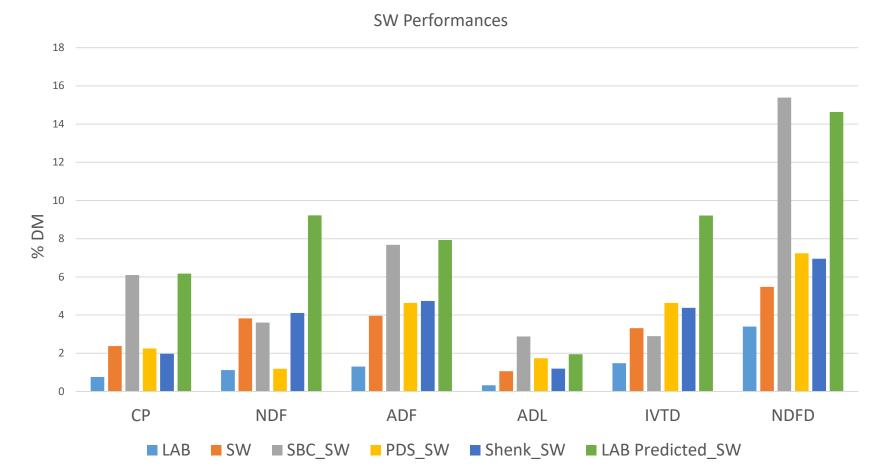


Wavelength (nm)



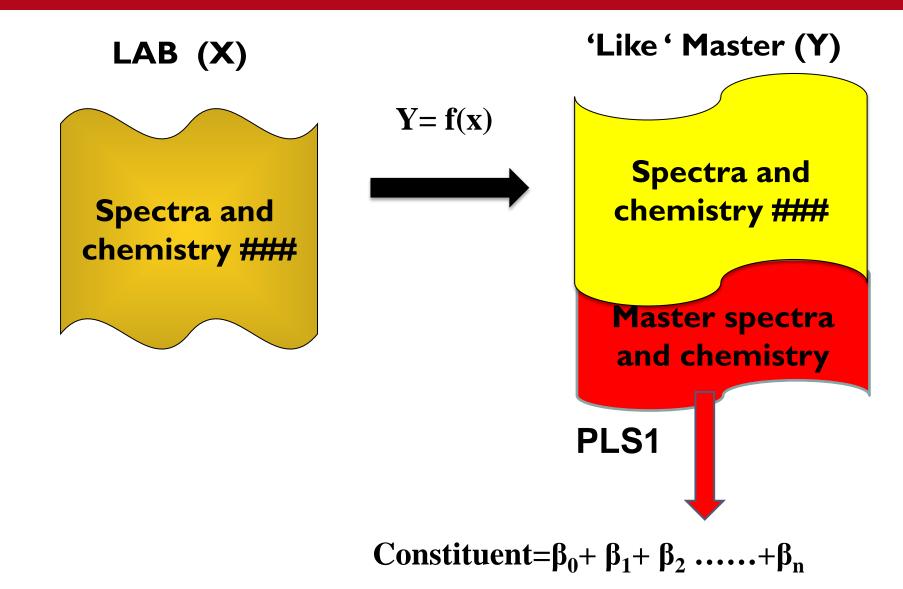
SW SEPs Alfalfa

PDS (+8%) < S&W (+16%) <SBC (+92%) < LAB_pred (+145%)





Calibration transfer -Standardization + updates





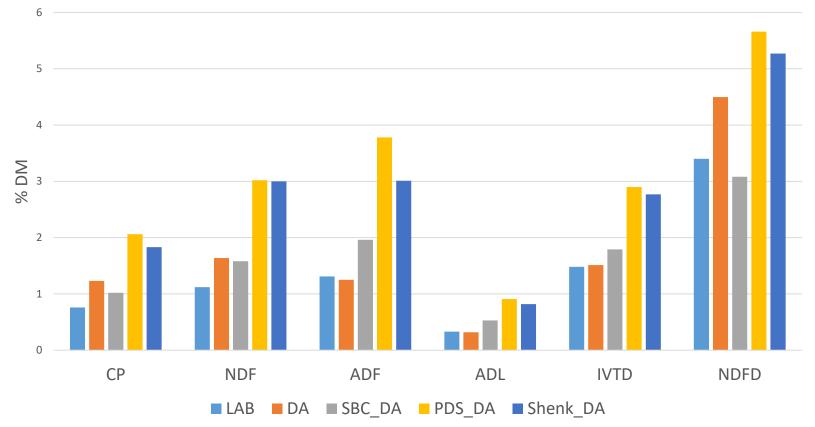
Università degli Studi

DI PADOVA

DA SEPs Alfalfa + Inoculation

SBC (-5%) < S&W (+60%) <PDS (+75%)







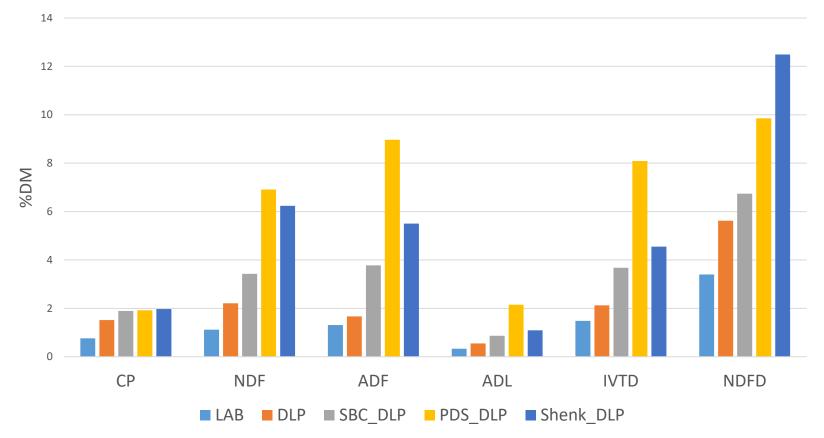
Università degli Studi

DI PADOVA

DLP SEPs Alfalfa + Inoculation

SBC (+50%) < S&W (+132%) <PDS (+175%)

DLP Performance + Inoculation

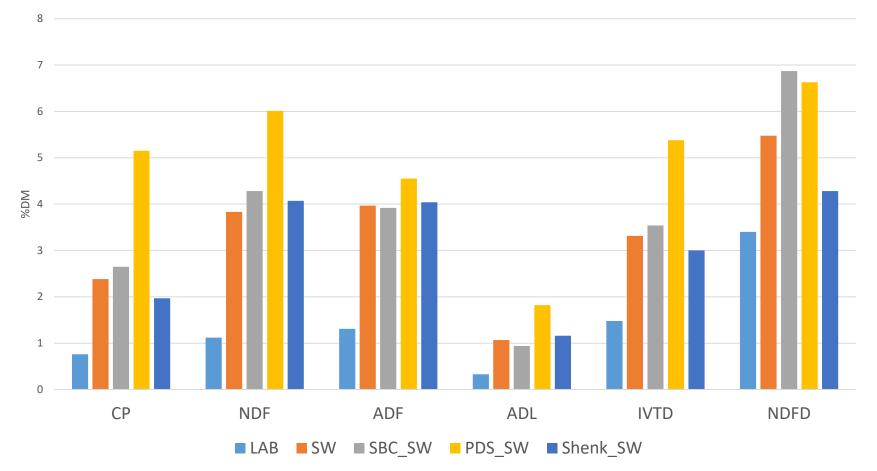




SW SEPs Alfalfa + Inoculation

SBC (+11%) < S&W (+90%) <PDS (+147%)

SW Performance + Inoculation





Conclusions



Must use many bullets to kill the beast

- Spectra correction: sometime the most sophisticated method is not the best choice
- Inoculation: highly effective, always have this options included
- Reduce calibration sensistivity (fewer PLS component, orthogonalization, REP file....)
- Bias adjustment of results may at the end be necessary



Conclusions

- Clearly one method would not fit all possible sensors
- Must have a high degree in chemometric to evaluate all possible solutions
 - looking for the future:
 - Easier and more effective methods of spectra transfer (Chemometric for Paolo)
 - May be an intelligent system that would make all possible evaluation



UNIVERSITÀ Plans for the Future

learn from mistakes

DI PADOVA

strive for improvements...

....to make cows and farmers happy





1222·2022 A N N I



Università degli Studi di Padova