



How to Analyse, Structure & Handle NIR Results in R Resiliently

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https://github.com/zoltankovacs/NIR2025-HASH_NIR4

SUNDAY JUNE 8, 2025

10:00-12:00

Sala dell'Editoria

Precourse 2

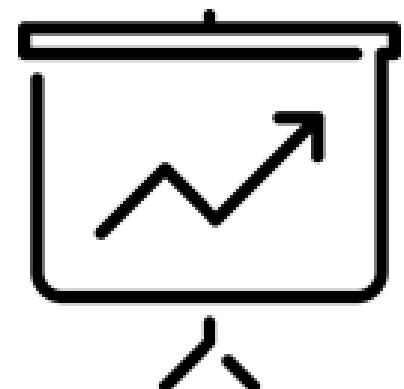
#NIR4 - How to Analyse, Structure & Handle NIR Results in R Resiliently

Outline

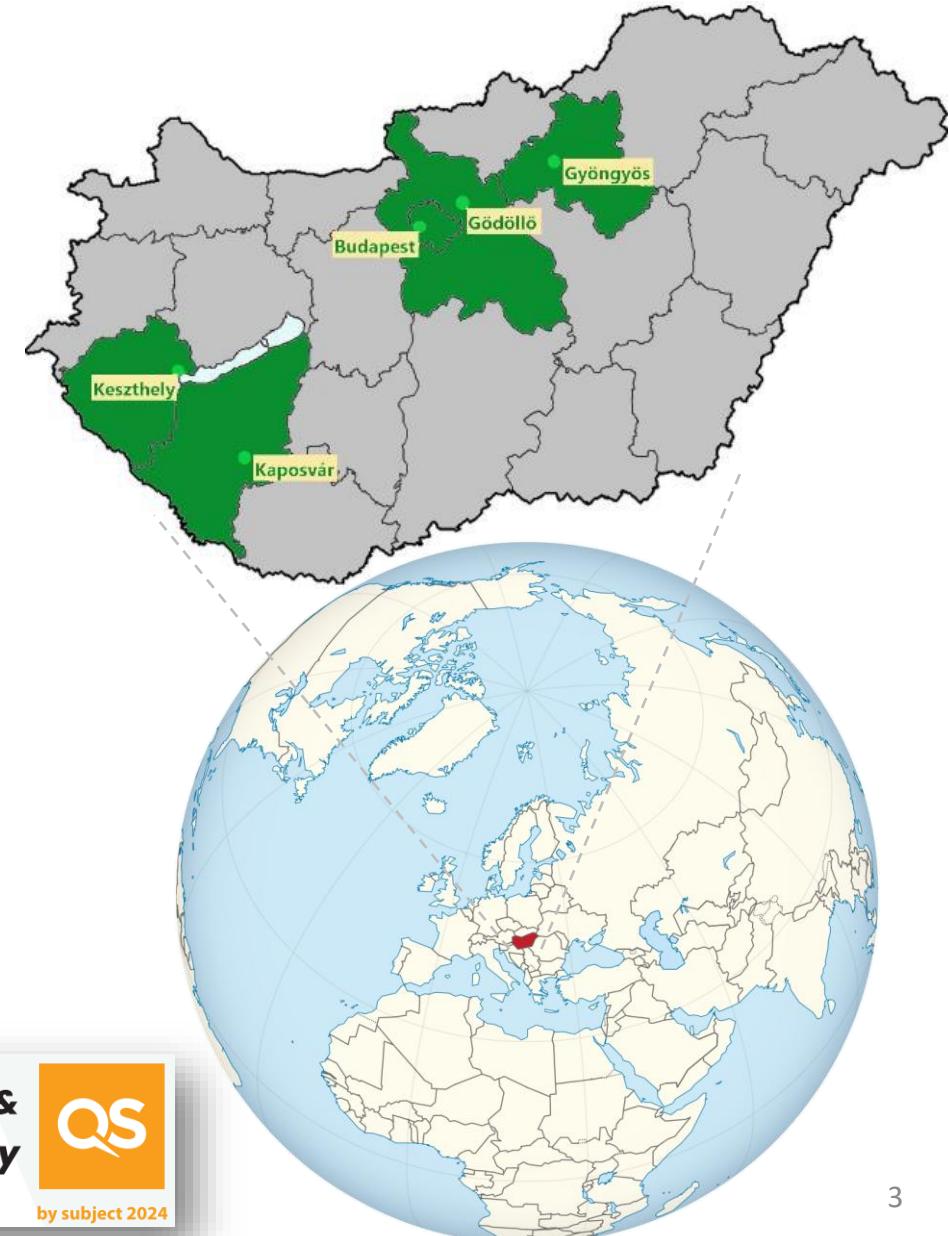
How to Analyse, Structure & Handle NIR Results in R Resiliently



- Brief self introduction
- Importance of NIRS in recent agriculture and food production
- Objective of this precourse
- An overview of aquap2
- Case study: adding data to a sample NIR – R database
- Aquap2 practice



- **Founded on 1 February 2021** (successor to former Szent Istvan University, Kaposvar University and National Agricultural Research and Innovation Centre)
- **One of the largest agricultural-focused, multi-disciplinary higher education institutions in CEE**
- **provides world-class education in agriculture-related fields**
- **No. of institutes: 21**
- **No. of students: 13 621**
- **No. of international students: 2076** (100+ countries) **15.24%**
- **Number of PhD schools: 12**
- **Number of PhD students: 864** (international PhD students 382) **44.2%**
- **Number of academic staff: 1060**
- **Number of staff: 1412**
- **Languages of instruction: Hu, Eng**



MATE Campuses and Buda Campus Institutes

Hungarian University of Agriculture and Life Sciences:

– Szent Istvan Campus, Gödöllő – Head Office

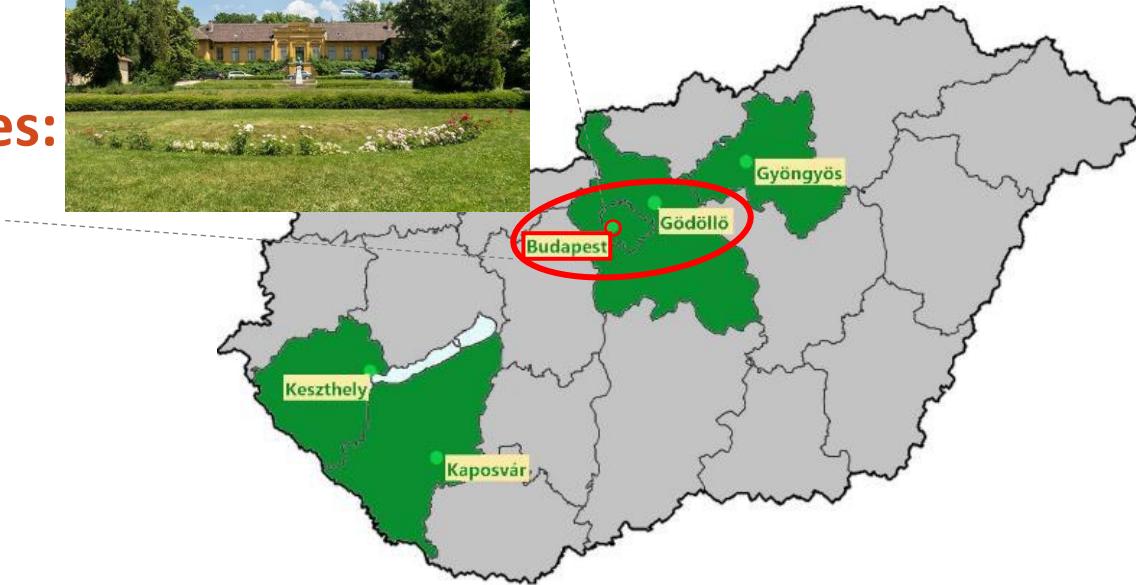
– Buda Campus

- Institute of Food Science and Technology
- Institute of Horticultural Sciences
- Institute of Landscape Architecture, Urban Planning and Garden Art
- Institute of Plant Protection
- Institute for Viticulture and Oenology

– Kaposvár Campus

– Gyöngyös Campus

– Keszthely Campus



Across the entire Food Chain:

production of food industry raw materials – **food processing** – food trading

Food processing technologies

- Animal product technologies (meat, milk, fish, egg):
 - Modern meat preservation and processing techniques etc.
 - Modern technologies, products, special diet, functional food etc.
- Fermentation technologies:
 - Beer, spirits, fermented functional food, pro/pre/synbiotics etc.
- Cereal science:
 - Modern bakery products, smart products etc.
- Fruit and vegetable processing technologies:
 - Drying technologies, aseptic technologies, PEF technology, etc.
- Modern postharvest and preservation technologies
 - Packaging, cold chain, minimal processing technologies (HHP, ultrasound)
 - Smart fresh technologies, etc.



Institute of Food Science and Technology

Research topics in focus II.

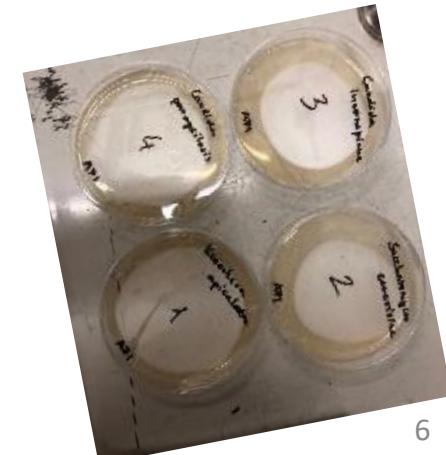
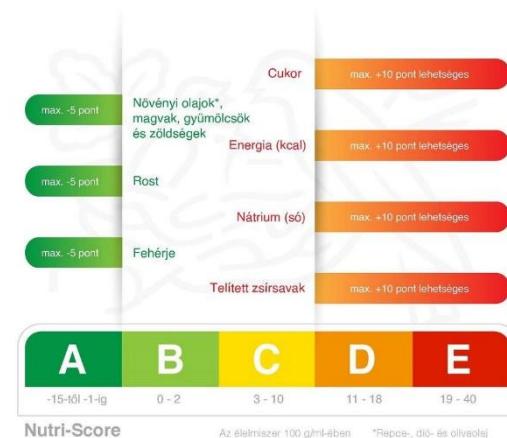
Across the entire Food Chain:

production of food industry raw materials – **food processing** – **food trading**

Food safety, quality, nutrition and trading



- Food safety:
 - Food microbiology, food analytics and food chemistry, microplastics, toxins, pesticide residues etc.
- Nutrition
 - Nutritional value analysis
 - Biological value enhancement technologies
- Food trading:
 - Consumer acceptance and preferences, logistics, marketing and trading
- *State of the art* research methodologies and instruments:
 - Analytical laboratories, sensory laboratory etc.



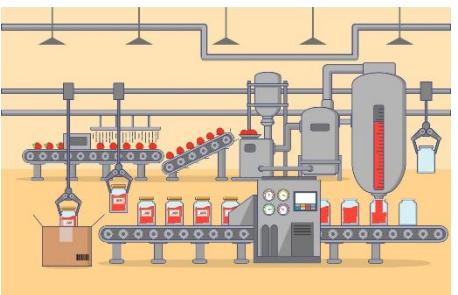
Institute of Food Science and Technology

Research topics in focus III.

Across the entire Food Chain:

production of food industry raw materials – **food processing – food trading**

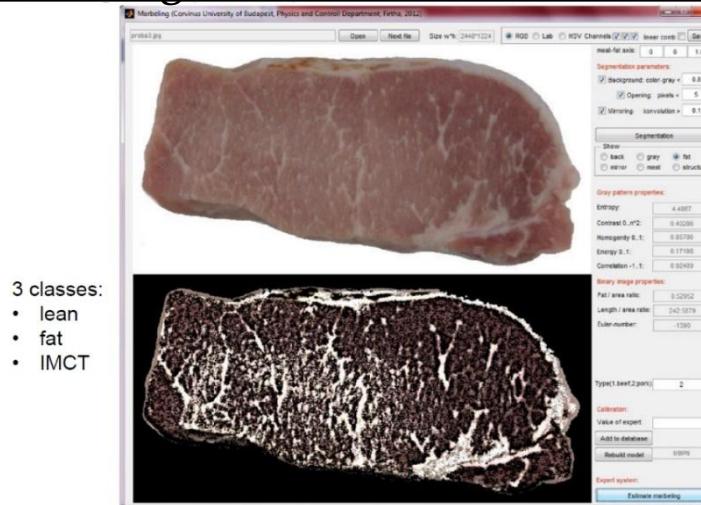
Measurement techniques, automatization, digitalization



- Robotization
- Data processing and evaluation, bigdata, chemometrics
- Digitalization, industry 4.0, IoT, M2M
- Modern measurement methods:
 - **Non-destructive methods**, contactless measurement technologies
 - Instrumental taste and odor sensing, **NIR**, hyperspectral, machine vision, etc.



Meat marbling measurement- machine vision



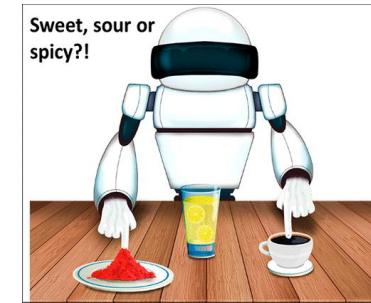
Quality assessment of food - NIRS and aquaphotonics

- | | |
|--|---|
|  Honey
Adulterated with rice syrup, self-produced sugar syrup and high fructose content sugar syrup |  Coffee drinks
Classification and quantification of different origin Arabica and Robusta coffee |
|  Meat mixture extracts
Detection and quantification of pork in beef |  Yogurt quality
With different lactobacillus strains grown with different water |
|  Mung bean juice
Monitoring of germination and prediction of ascorbic acid |  Tokaj wine
Adulterated with grape must concentrate |
|  Monilia (brown rot) on plums
Early phase detection of <i>Monilia fructigena</i> infection on plums |  Tomato concentrate
Adulterated with starch, paprika seed powder, NaCl and sucrose |
|  Paprika powder dilution
Adulterated with corn flour at different concentration levels |  Mineral water
With different mineral content mixed and analyzed |

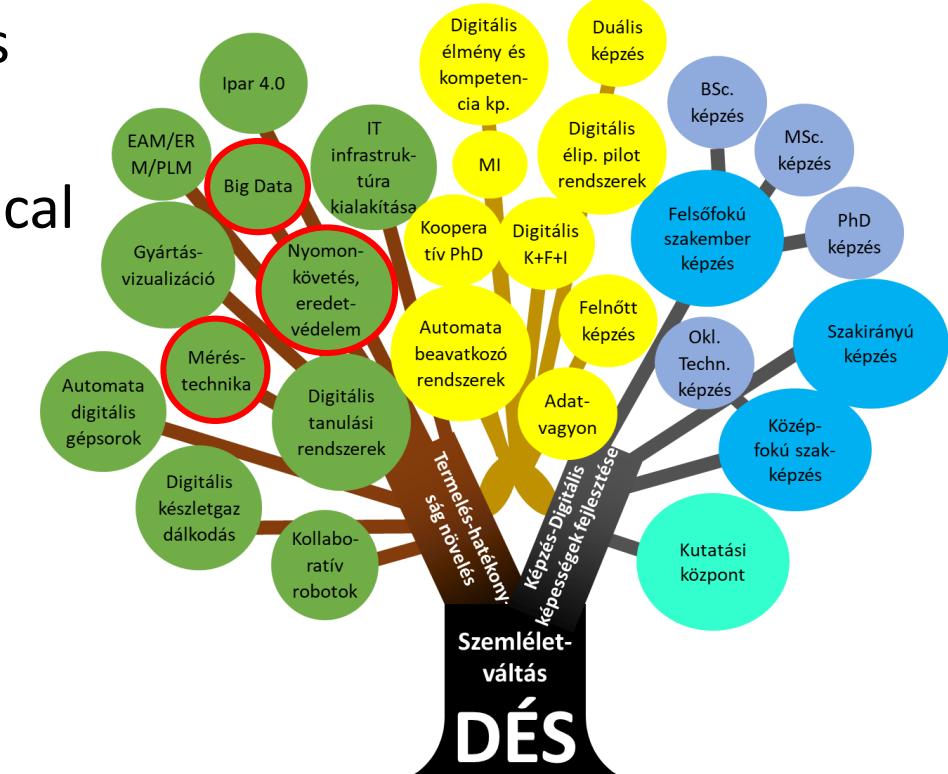


Importance of the non-destructive analytical techniques

- Precision farming and food production
- Increasing quantitative needs ←→ the standard quality food ←→ more economically
- Data driven decision making
- Rapid and precise quality determination of raw materials and food products
- Immense need for advanced and non-destructive analytical techniques in the agriculture and food industry:
 - ultrasonic measurement,
 - bioelectric properties, impedance spectroscopy,
 - acoustic measurement,
 - electronic tongue and electronic nose
 - machine vision,
 - near-infrared spectroscopy and aquaphotomics etc.

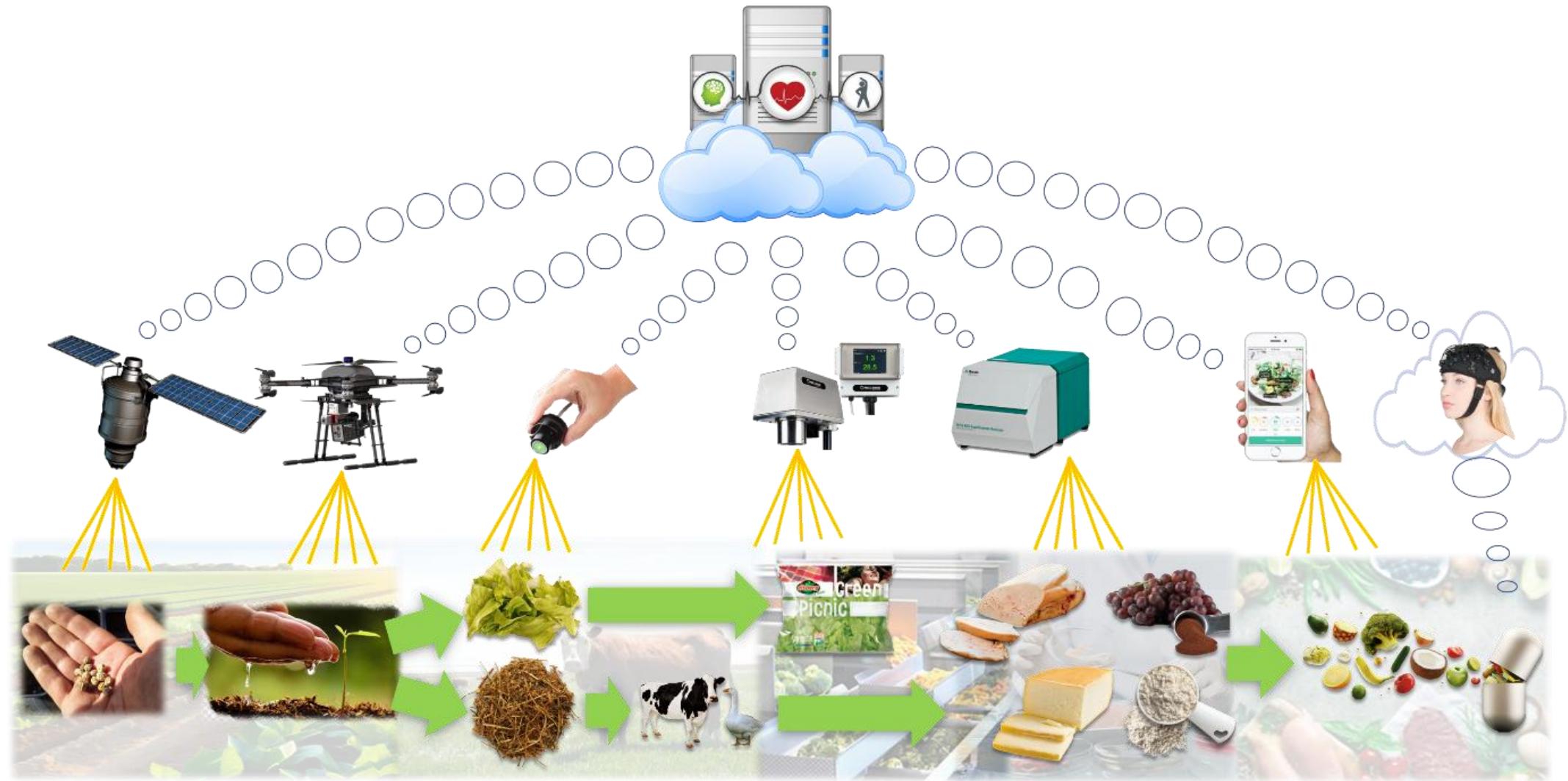


ACS Sens. 2018, 3, 11, 2375–2384



Objective of this presentation

to provide insights about recent results of NIRS and aquaphotomics supporting the monitoring of food supply chain “from farm to fork” – and beyond.





Food quality assessment with NIR and aquaphotomics



Lettuce quality

Monitoring of lettuce during storage in argon gas packaging



Goose Liver quality

Determination of the blood content in fattened goose liver



Honey authentication

Detection of heat treatment and syrup adulteration of honeys



Fruit Juice Fortification

Understanding water structure changes in fruit juices induced by plant extracts using aquaphotomics



Monilia (brown rot) on plums

Early phase detection of *Monilia fructigena* infection on plums



Coffee drinks

Classification and quantification of different origin Arabica and Robusta coffee



Meat mixture extracts

Detection and quantification of pork in beef



Tokaji wine

Adulterated with grape must concentrate



Mung bean juice

Monitoring of germination and prediction of ascorbic acid



Mineral water

With different mineral content mixed and analyzed



Enriched cheese quality

Impact of Treatment and Aging on PUFA-Enriched Cheese: an Aquaphotomics focused approach



Enriched cheese quality

Comparison of regression models for the prediction of caffeine content in aqueous solutions + ANN

Objectives of this course



<https://www.inno-spectra.com>

to introduce some possible applications of R project and the aquap2 package

- to design experiments,
- to structure, curate and archive spectral data effectively

All in all, to present hands-on practical examples – from experimental design to saving and structuring the collected data as part of a larger NIR database.





An overview of aquap2

Multivariate analysis tools for aquaphotonics



Data analysis with the “aquap2” R-package: a free multivariate data analysis framework



<https://github.com/bpollner/aquap2>

https://github.com/zoltankovacs/NIR2025-HASH_NIR4



Package "aquap2"

```
## the universal input, checking for the class of the input-object
pickPeaks <- function(ObjectToPickPeaks, bandwidth=25, comps=1:4, discrim=FALSE) {      ### universal peak picker / extracto
  if (class(ObjectToPickPeaks) == "mvr") {
    allColNames <- colnames(ObjectToPickPeaks$coefficients)
    lastName <- allColNames[length(allColNames)]
    if (!is.character(lastName)) {      ## problem if there is only one component -- we do not get a name back then ....
      lastName <- "1 comps"
    }
    mat <- ObjectToPickPeaks$coefficients[, , ObjectToPickPeaks$ncomp]
    dfToPickPeaks <- data.frame(X=mat)
    colnames(dfToPickPeaks) <- lastName
  }
  else {                                ## for SVD or PCA objects
    mat <- ObjectToPickPeaks$loadings[, , ObjectToPickPeaks$ncomp]
    dfToPickPeaks <- data.frame(X=mat)
    colnames(dfToPickPeaks) <- ObjectToPickPeaks$loadings[, , ObjectToPickPeaks$ncomp]
  }
  if (class(ObjectToPickPeaks) == "data.frame") {
    dfToPickPeaks <- ObjectToPickPeaks
  }
}
pickPeaks <- function(pickResults, bandwidth, discrim)
} # EOF
```

- drastically speeds up analysis time
- highly repetitive tasks get completely scriptable (i.e., automated)

Key Features

Experiment Design

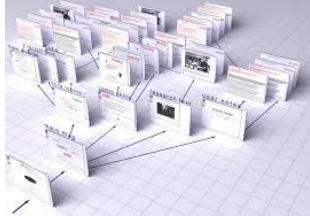
- truly randomize samples
- facilitates time resolved experiments

Data Import & Organization

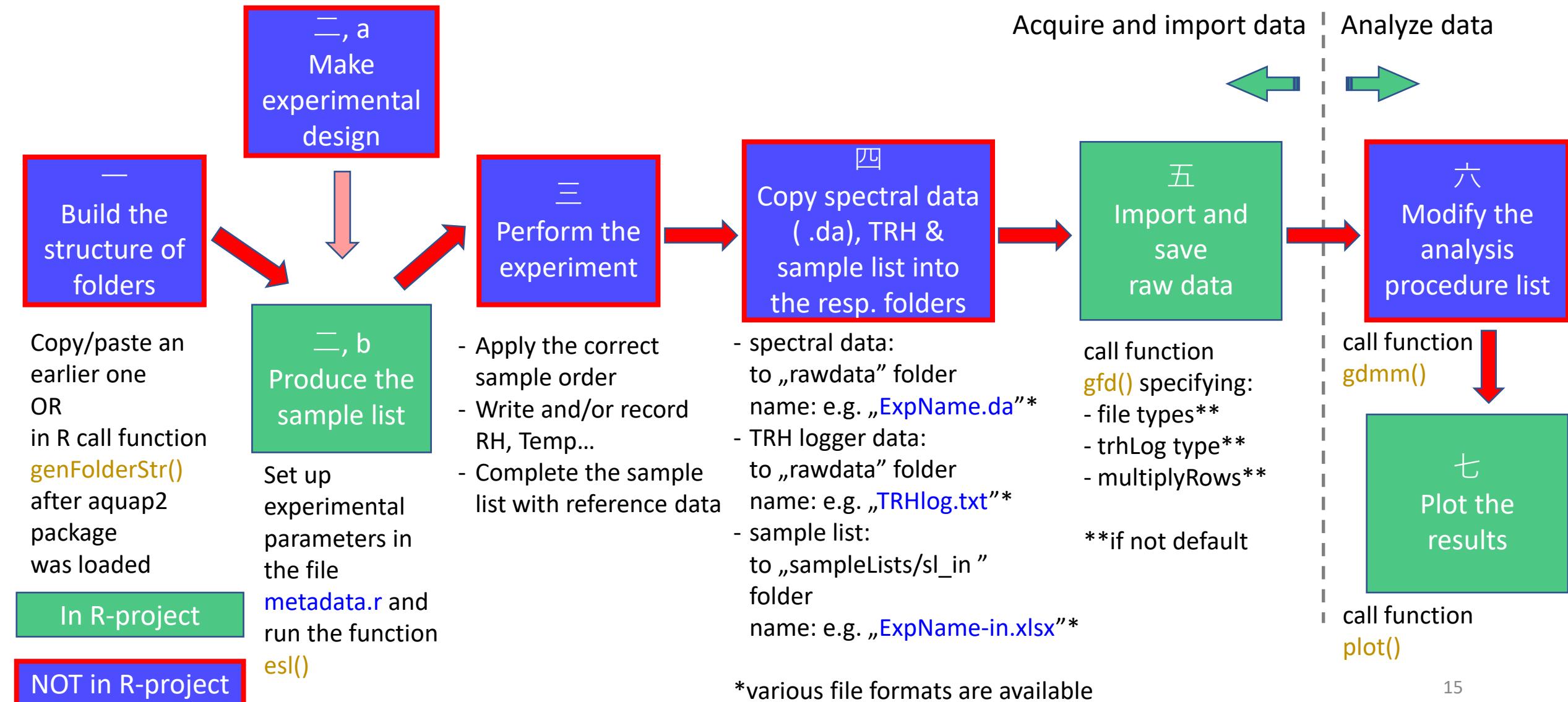
- import of metadata: class.- and numerical variables
- import of spectral data
- consistent coloring
- align environmental data (T, rH) to timestamp on spectra
- add custom, device-specific noise to dataset

Data Analysis & Visualization

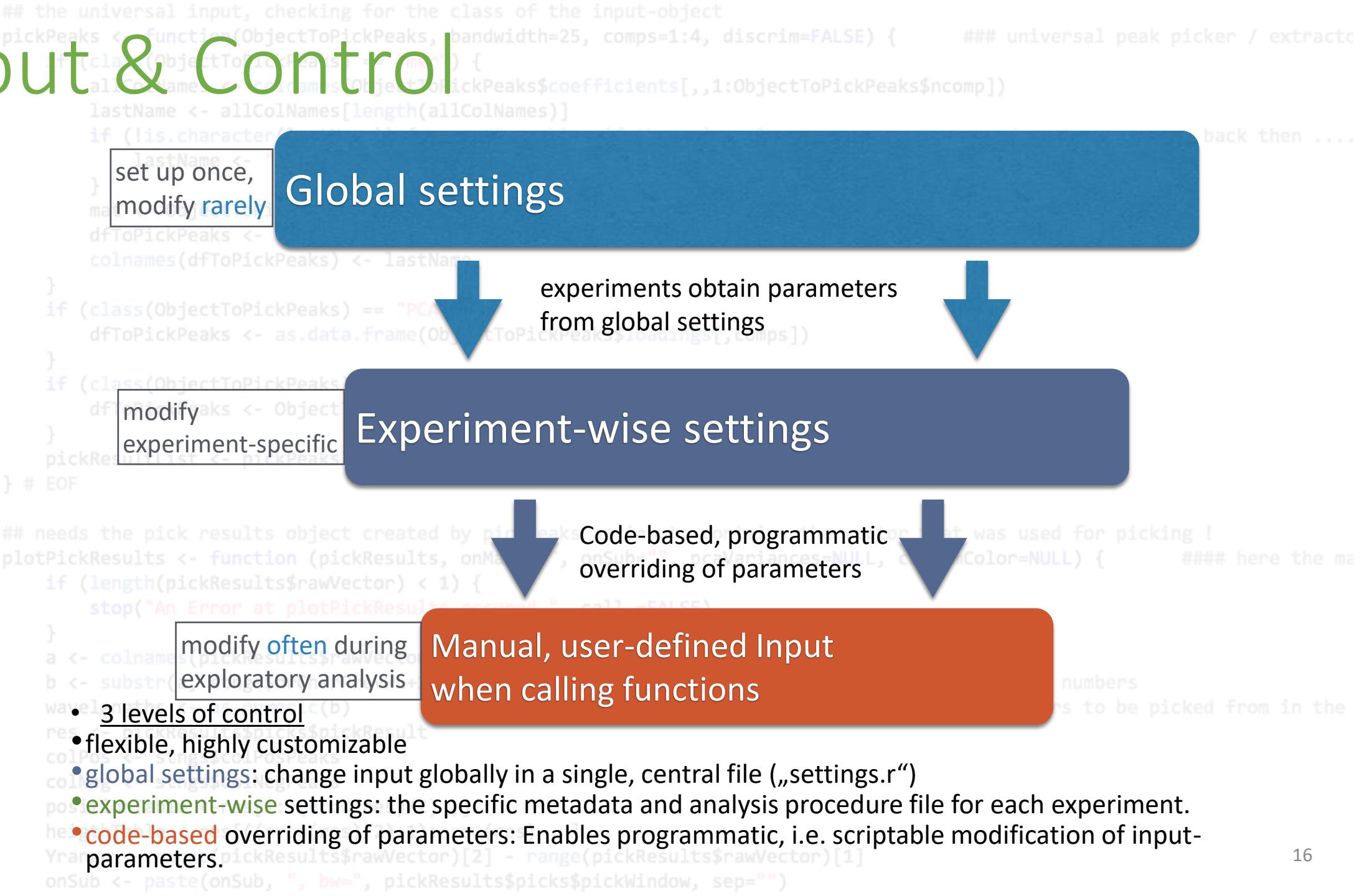
- User-defined splitting of dataset
- **Bootstrapped** Aquagram: custom temperature calibration data
- **User-defined splitting of dataset by wavelength or any class-variable**
- **Highly flexible input and control system**



Workflow of the “aquap2” package



Input & Control



Measurement preparation

- ✓ Generate the most important folders and files
- ✓ Create and export samples list based on „metadata.R”

```
## the universal input, checking for the class of the input-object
pickPeaks <- function(ObjectToPickPeaks, bandwidth=25, comps=1:4, discrim=FALSE) {      ### universal peak picker / extracto
  allColNames <- colnames(ObjectToPickPeaks$coefs$coefficients[, , 1:ObjectToPickPeaks$ncomp])
  lastName <- allColNames[length(allColNames)]
  if (!is.character(lastName)) { ## problem if there is only one component - we do not get a name back then ....
    lastName <- "ObjectToPickPeaks$coefs$coefficients[, , 1:ObjectToPickPeaks$ncomp]"
  }
  colnames(dfToPickPeaks) <- lastName
}

17 TimePoints <- FALSE          ## how many points in time does the experiment cover? Provide a label for each sample
18 nrConScans <- 3             ## how many consecutive scans for every sample?
19 spacing <- FALSE            ## the space between environmental control samples (set to FALSE for not inserting)
20 #####
21 columnNamesL1 <- c("C_Water", "Y_waterTemp")      ## please note that there is a special prefix for class-
22 columnNamesL2 <- c("C_DELETE", "C_DELETE")        ## "DELETE" is the default character for those columns that
23 L1 <- list(list("MQ", "TAP", "DW", "AIR"), list(seq(20, 74, 2)))           ## Please look at the vignette
24 L2 <- L1
25 #####
26 Repls <- 1                  ## how many replications for each sample?
27 Group <- c("no")            ## additional groups to split the above generated classes into, like e.g. experimental
28
```

Metadata

```
  stop("An Error at plotPickResults occurred.", call.=FALSE)
}
a <- colnames(pickResults$rawVector)
b <- substr(a, stngs$nrCharPrevWL+1, nchar(a))      ## to get rid of the "w" in front of the numbers
wavelengths <- as.numeric(b)          # so we have the wavelength in the column, and the vectors to be picked from in the
res <- pickResults$picks$pickResult
colPos <- stngs$colPosPeaks
colNeg <- stngs$colNegPeaks
positionTable <- res[1: (nrow(res)/2) ,]
heightTable <- res[((nrow(res)/2)+1):nrow(res) ,]
Yrange <- range(pickResults$rawVector)[2] - range(pickResults$rawVector)[1]
onSub <- paste(onSub, ", bw=", pickResults$picks$pickWindow, sep="")
```

Measurement preparation

- ✓ Generate the metadata
- ✓ Create and export

```
17 TimePoints <- FALSE
18 nrConScans <- 3
19 spacing <- FALSE
20 #####
21 columnNamesL1 <- c("C_Water",
22 columnNamesL2 <- c("C_DELETE",
23 L1 <- list(list("MQ", "TAP",
24 L2 <- L1
25 #####
26 Repls <- 1
27 Group <- c("no")
28
```

Metadata

```
> # 1.) Creat sample list based on metadata file
> esl(form = "xlsx", rnd = FALSE, showFirstRows = TRUE, timeEstimate = TRUE)
Creating sample list...
Writing sample list to file...
A sample list in .xlsx format with 112 rows has been saved to "sampleLists/s1_out".
Minimum working time: 9.5 hours.
Have fun!
```

	Y_SampleNr	C_Time	C_ECRM	C_Water	Y_waterTemp	C_Repl	C_Group	Y_Temp	Y_RelHum
1	1	NT	RM	MQ	20	R1	no		
2	2	NT	RM	TAP	20	R1	no		
3	3	NT	RM	DW	20	R1	no		
4	4	NT	RM	AIR	20	R1	no		
5	5	NT	RM	MQ	22	R1	no		
6	6	NT	RM	TAP	22	R1	no		
7	7	NT	RM	DW	22	R1	no		
8	8	NT	RM	AIR	22	R1	no		
9	9	NT	RM	MQ	24	R1	no		
10	10	NT	RM	TAP	24	R1	no		
11	11	NT	RM	DW	24	R1	no		
12	12	NT	RM	AIR	24	R1	no		
13	13	NT	RM	MQ	26	R1	no		
14	14	NT	RM	TAP	26	R1	no		
15	15	NT	RM	DW	26	R1	no		
16	16	NT	RM	AIR	26	R1	no		
17	17	NT	RM	MQ	28	R1	no		
18	18	NT	RM	TAP	28	R1	no		
19	19	NT	RM	DW	28	R1	no		
20	20	NT	RM	AIR	28	R1	no		
21	21	NT	RM	MQ	30	R1	no		

Console

Data analysis preparation

- ✓ Specify the raw data details in „metadata.R”
- ✓ and import

```
## the universal input, checking for the class of the input-object
pickPeaks <- function(ObjectToPickPeaks, bandwidth=25, comps=1:4, discrim=FALSE) {      ### universal peak picker / extracto
  allColNames <- colnames(ObjectToPickPeaks$coefficients[,1:ObjectToPickPeaks$ncomp])
  lastName <- allColNames[length(allColNames)]
  if (!is.character(lastName)) {
    # problem if there is only one component -- we won't get a name back then ....
    lastName <- paste("comp", 1)
  }
  mat <- ObjectToPickPeaks$coefficients[, ObjectToPickPeaks$ncomp]
  dFToPickPeaks <- data.frame(X=mat)
  colnames(dFToPickPeaks) <- lastName
}

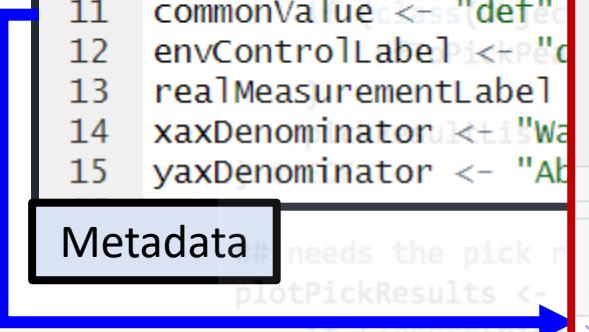
# 1.) Set up the experiment
expName <- "TempData"
## the name of the experiment
filetype <- "vision_NIR"
noiseFileName <- "def"
tempCalibFileName <- "def"
commonValue <- "def"
envControlLabel <- "def"
realMeasurementLabel <- "def"
xaxDenominator <- "Wavelength"
yaxDenominator <- "Absorbance"
```

Script

```
7 expName <- "TempData"
8 filetype <- "vision_NIR"
9 noiseFileName <- "def"
10 tempCalibFileName <- "def"
11 commonValue <- "def"
12 envControlLabel <- "def"
13 realMeasurementLabel <- "def"
14 xaxDenominator <- "Wavelength"
15 yaxDenominator <- "Absorbance"

## the name of the experiment
32 # 2.) Import data after experiment and copying all files on place
33 fullData <- gfd(ttl = FALSE) # every detail of the spectral data is defined in the "metadata.R" file
34
35 # The complete dataset is saved in the "R-data" folder and can be read in simply by running:
36 fullData <- gfd()
37
38
39
40
```

Metadata



Console

```
R - R 4.5.0 - C:/Users/vit9178/Downloads/aquap2_course_Japan2025-main/aquap2_course_Rome2023@home/
> # 2.) Import data after experiment and copying all files on place
> fullData <- gfd(ttl = FALSE) # every detail of the spectral data is defined in the "metadata.R" file
Importing data...
  detecting outliers... found *81*. [Using 1 components.]
  Aligning temp. and rel.hum. values to timestamp...
  Done.
Dataset saved under "R-data/TempData".
>
>
>
> # The complete dataset is saved in the "R-data" folder and can be read in simply by running:
> fullData <- gfd()
Dataset "TempData" was loaded.
```

MVA Methods – close look at the „anproc.R” file

Already Implemented

- Data pre-treatment
 - smoothing, 2nd derivative, SNV, MSC, EMSC, deTrend, gap derivative, ...
- PCA; SIMCA; PLSR
- DA
- Aquagram (classic & extended)

In development

- PLS-DA; ANN; SVM; ICA; KNN

Key spectral prepoceessing

Data splitting

- By grouping variables,
- By wavelength ranges...

Data treatment

- Specific spectral processing,
- Averaging,
- Outlier elimination...

Anproc	
10	sp1.var <- NULL
11	sp1.wl <- NULL
12	#####
13	dpt.pre <- NULL
14	#####
15	sp1.do.csAvg <- FALSE
16	sp1.csAvg.raw <- TRUE
17	#
18	sp1.do.noise <- FALSE
19	sp1.noise.raw <- TRUE
20	#
21	sp1.do.exOut <- FALSE
22	sp1.exOut.raw <- FALSE
23	sp1.exOut.var <- c("C_FooBar")
24	#####
25	dpt.post <- NULL

by which variables should the dataset be split?
which wavelengths to use? leave empty c() or set
Character vector, which of the available modules
if all the consecutive scans of a single sample
if, should the consecutive scans of a single s
if artificial noise should be added to the dat
if, should the noise-test be performed, the ran
if exclusion of outliers should be performed
if, should exclusion of outliers be performe
A vector with valid class variable names
Character vector, which of the available modules

User-defined splitting of dataset

C_Time	C_Group	w1300	w1300.5	w1301	w...	w1600
T0	Treatment_AAA	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_AAA	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_AAA	0.2362	0.2345	0.4352	0.2362	0.2362
T0	Treatment_BBB	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_BBB	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_BBB	0.2362	0.2345	0.4352	0.2362	0.2362
T0	Treatment_CCC	0.2345	0.4352	0.2362	0.4352	0.2345
T0	Treatment_CCC	0.4352	0.2362	0.2345	0.2362	0.4352
T0	Treatment_CCC	0.2362	0.4352	0.4352	0.2345	0.2362
T1	Treatment_AAA	0.2345	0.2362	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.4352	0.2345	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.2362	0.4352	0.2345	0.2362	0.2345
T1	Treatment_BBB	0.2345	0.2362	0.4352	0.2362	0.4352
T1	Treatment_BBB	0.4352	0.2345	0.2362	0.2345	0.2362
T1	Treatment_BBB	0.2362	0.4352	0.2345	0.4352	0.2362
T1	Treatment_CCC	0.4352	0.2362	0.4352	0.2362	0.2345
T1	Treatment_CCC	0.2362	0.2345	0.4352	0.2362	0.2362
T1	Treatment_CCC	0.2345	0.4352	0.2362	0.2345	0.2345

Key Features II: User-defined splitting of dataset

Split by „C_Time“

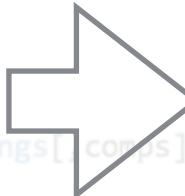
C_Time	C_Group	w1300	w1300.5	w1301	w...	w1600
T0	Treatment_AAA	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_AAA	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_AAA	0.2362	0.2345	0.4352	0.2362	0.2362
T0	Treatment_BBB	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_BBB	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_BBB	0.2362	0.2345	0.4352	0.2362	0.2362
T0	Treatment_CCC	0.2345	0.4352	0.2362	0.4352	0.2345
T0	Treatment_CCC	0.4352	0.2362	0.2345	0.2362	0.4352
T0	Treatment_CCC	0.2362	0.4352	0.4352	0.2345	0.2362
T1	Treatment_AAA	0.2345	0.2362	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.4352	0.2345	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.2362	0.4352	0.2345	0.2362	0.2345
T1	Treatment_BBB	0.2345	0.2362	0.4352	0.2362	0.4352
T1	Treatment_BBB	0.4352	0.2345	0.2362	0.2345	0.2362
T1	Treatment_BBB	0.2362	0.4352	0.2345	0.4352	0.2362
T1	Treatment_CCC	0.4352	0.2362	0.4352	0.2362	0.2345
T1	Treatment_CCC	0.2362	0.2345	0.4352	0.2362	0.2362
T1	Treatment_CCC	0.2345	0.4352	0.2362	0.2345	0.2345

Key Features II: User-defined splitting of dataset

```
## the universal input, checking for the class of the input-object
pickPeaks <- function(ObjectToPickPeaks, bandwidth=25, comps=1:4, discrim=FALSE) {
  ## universal peak picker / extractor
  if (is.data.frame(ObjectToPickPeaks)) == TRUE {
    allColNames <- colnames(ObjectToPickPeaks)
    pickPeaks <- pickPeaksInner(dfToPickPeaks, bandwidth, discrim)
    lastName <- allColNames[length(allColNames)]
    if (!is.character(lastName)) { ## problem if there is only one component -- we do not get a name back then ....
      lastName <- "1_comps"
    }
  } # if (is.data.frame)
  if (is.matrix(ObjectToPickPeaks)) == TRUE {
    pickPeaks <- pickPeaksInner(dfToPickPeaks, bandwidth, discrim)
  }
  if (is.list(ObjectToPickPeaks)) == TRUE {
    pickPeaks <- pickPeaksInner(dfToPickPeaks, bandwidth, discrim)
  }
}
} # EOF
```

C_Time	C_Group	w1300	w1300.5	w1301	w...	w1600
T0	Treatment_AAA	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_AAA	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_AAA	0.2362	0.2345	0.4352	0.2362	0.2362
T0	Treatment_BBB	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_BBB	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_BBB	0.2362	0.2345	0.4352	0.2362	0.2362
T0	Treatment_CCC	0.2345	0.4352	0.2362	0.4352	0.2345
T0	Treatment_CCC	0.4352	0.2362	0.2345	0.2362	0.4352
T0	Treatment_CCC	0.2362	0.4352	0.4352	0.2345	0.2362

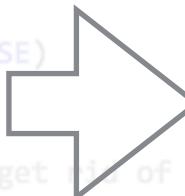
Peaks\$ncomp]



Separate Analysis and
Visualization

T1	Treatment_AAA	0.2345	0.2362	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.4352	0.2345	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.2362	0.4352	0.2345	0.2362	0.2345
T1	Treatment_BBB	0.2345	0.2362	0.4352	0.2362	0.4352
T1	Treatment_BBB	0.4352	0.2345	0.2362	0.2345	0.2362
T1	Treatment_BBB	0.2362	0.4352	0.2345	0.4352	0.2362
T1	Treatment_CCC	0.4352	0.2362	0.4352	0.2362	0.2345
T1	Treatment_CCC	0.2362	0.2345	0.4352	0.2362	0.2362
T1	Treatment_CCC	0.2345	0.4352	0.2362	0.2345	0.2345

```
t; contains the vector that was used for picking !
, pcaVariances=NULL, customColor=NULL) { ##### here the ma
SE)
get rid of the "w" in front of the numbers
length in the column, and the vectors to be picked from in the
```



```
colNeg <- stngs$colNegPeaks
positionTable <- res[1: (nrow(res)/2) ,]
heightTable <- res[((nrow(res)/2)+1):nrow(res) , ]
Yrange <- range(pickResults$rawVector)[2] - range(pickResults$rawVector)[1]
onSub <- paste(onSub, ", bw=", pickResults$picks$pickWindow, sep="")
```

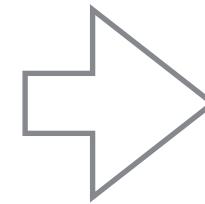
Key Features II: User-defined splitting of dataset

Split by „C_Group“

C_Time	C_Group	w1300	w1300.5	w1301	w...	w1600
T0	Treatment_AAA	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_AAA	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_AAA	0.2362	0.2345	0.4352	0.2362	0.2362
T0	Treatment_BBB	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_BBB	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_BBB	0.2362	0.2345	0.4352	0.2362	0.2362
T0	Treatment_CCC	0.2345	0.4352	0.2362	0.4352	0.2345
T0	Treatment_CCC	0.4352	0.2362	0.2345	0.2362	0.4352
T0	Treatment_CCC	0.2362	0.4352	0.4352	0.2345	0.2362
T1	Treatment_AAA	0.2345	0.2362	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.4352	0.2345	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.2362	0.4352	0.2345	0.2362	0.2345
T1	Treatment_BBB	0.2345	0.2362	0.4352	0.2362	0.4352
T1	Treatment_BBB	0.4352	0.2345	0.2362	0.2345	0.2362
T1	Treatment_BBB	0.2362	0.4352	0.2345	0.4352	0.2362
T1	Treatment_CCC	0.4352	0.2362	0.4352	0.2362	0.2345
T1	Treatment_CCC	0.2362	0.2345	0.4352	0.2362	0.2362
T1	Treatment_CCC	0.2345	0.4352	0.2362	0.2345	0.2345

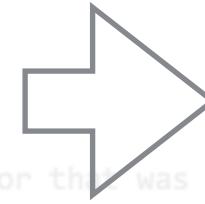
Key Features II: User-defined splitting of dataset

T0	Treatment_AAA	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_AAA	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_AAA	0.2362	0.2345	0.4352	0.2362	0.2362
T1	Treatment_AAA	0.2345	0.2362	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.4352	0.2345	0.2362	0.4352	0.2362
T1	Treatment_AAA	0.2362	0.4352	0.2345	0.2362	0.2345



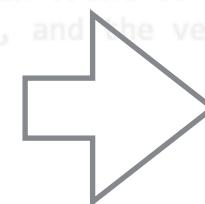
Separate Analysis and Visualization

T0	Treatment_BBB	0.2345	0.4352	0.2362	0.2345	0.2345
T0	Treatment_BBB	0.4352	0.2362	0.2345	0.4352	0.4352
T0	Treatment_BBB	0.2362	0.2345	0.4352	0.2362	0.2362
T1	Treatment_BBB	0.2345	0.2362	0.4352	0.2362	0.4352
T1	Treatment_BBB	0.4352	0.2345	0.2362	0.2345	0.2362
T1	Treatment_BBB	0.2362	0.4352	0.2345	0.4352	0.2362



Separate Analysis and Visualization

T0	Treatment_CCC	0.2345	0.4352	0.2362	0.4352	0.2345
T0	Treatment_CCC	0.4352	0.2362	0.2345	0.2362	0.4352
T0	Treatment_CCC	0.2362	0.4352	0.4352	0.2345	0.2362
T1	Treatment_CCC	0.4352	0.2362	0.4352	0.2362	0.2345
T1	Treatment_CCC	0.2362	0.2345	0.4352	0.2362	0.2362
T1	Treatment_CCC	0.2345	0.4352	0.2362	0.2345	0.2345



Spectrum-based calculations – *raw spectra*

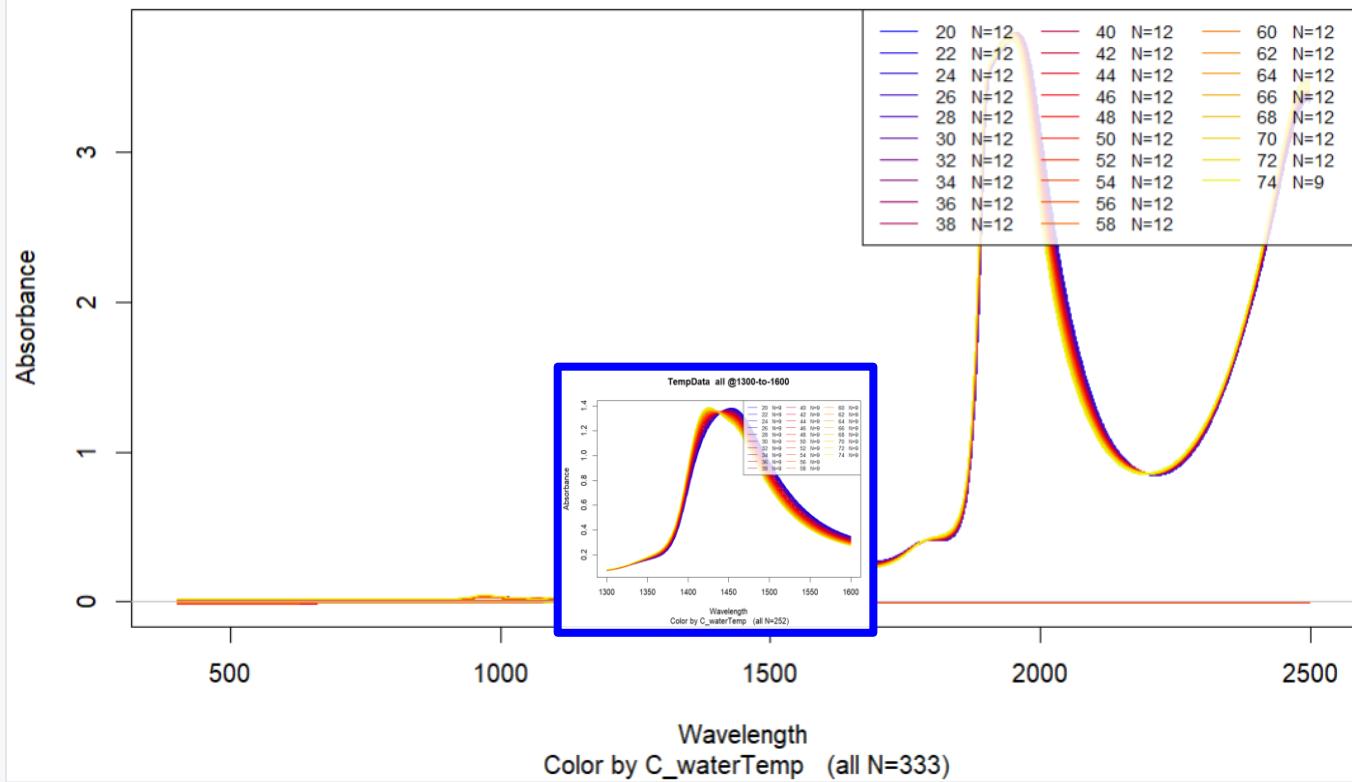
Script

```
37  
38  
39  
40 # 3.) Plot raw spectra from imported dataset  
41 plot(fullData, colorBy = "C_waterTemp")  
42 plot_spectra(fullData, colorBy = "C_waterTemp", pg.where = "")  
43  
44  
45
```

Console

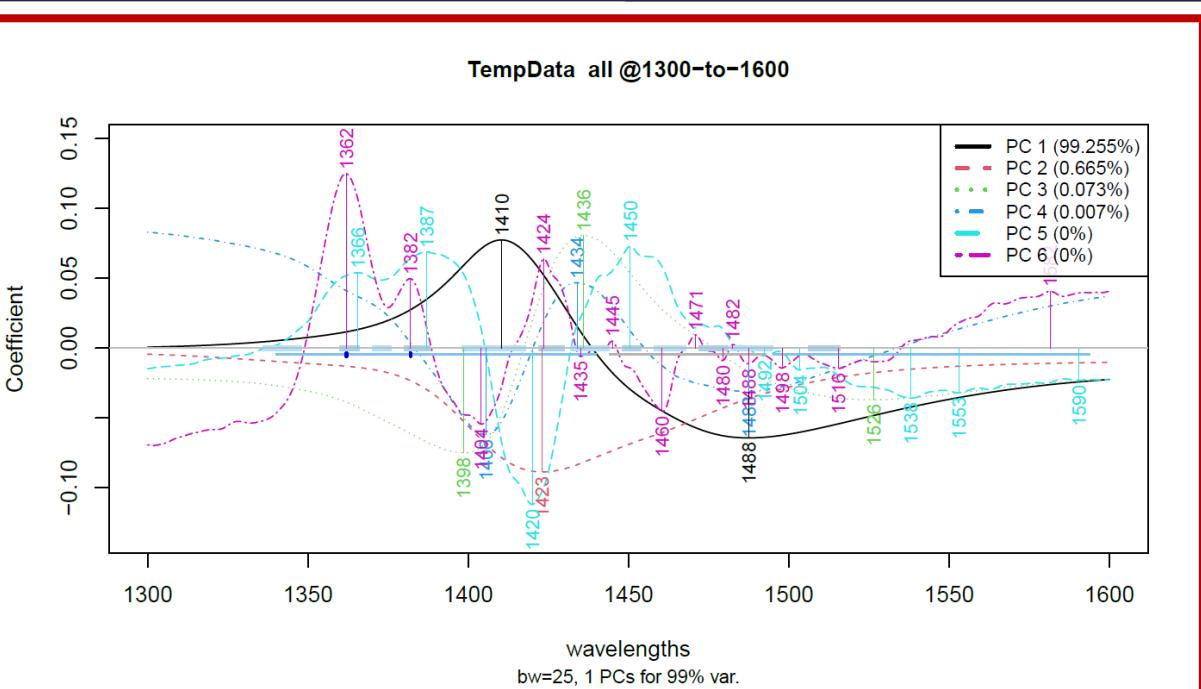
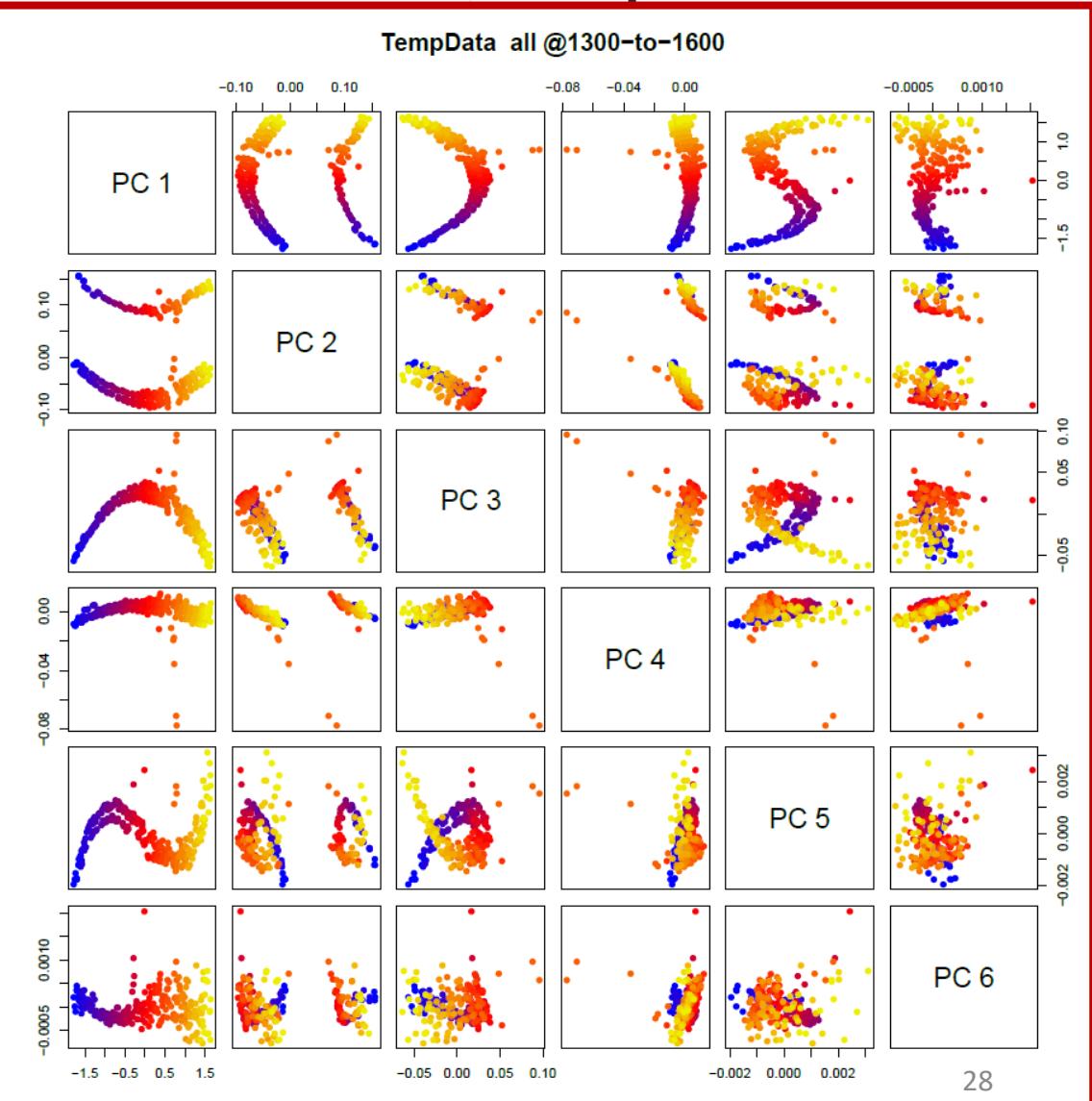
```
60:1 (Top Level)   
Console Terminal x Background Jobs x  
R 4.5.0 · C:/Users/vit9178/Downloads/aquap2_course_Japan2025-main/aquap2_course_Rome2023@home/  
> # 3.) Plot raw spectra from imported dataset  
> plot(fullData, colorBy = "C_waterTemp")  
Plotting raw spectra ... ok  
> plot_spectra(fullData, colorBy = "C_waterTemp", pg.where = "")  
> |
```

TempData



Spectrum-based calculations – PCA

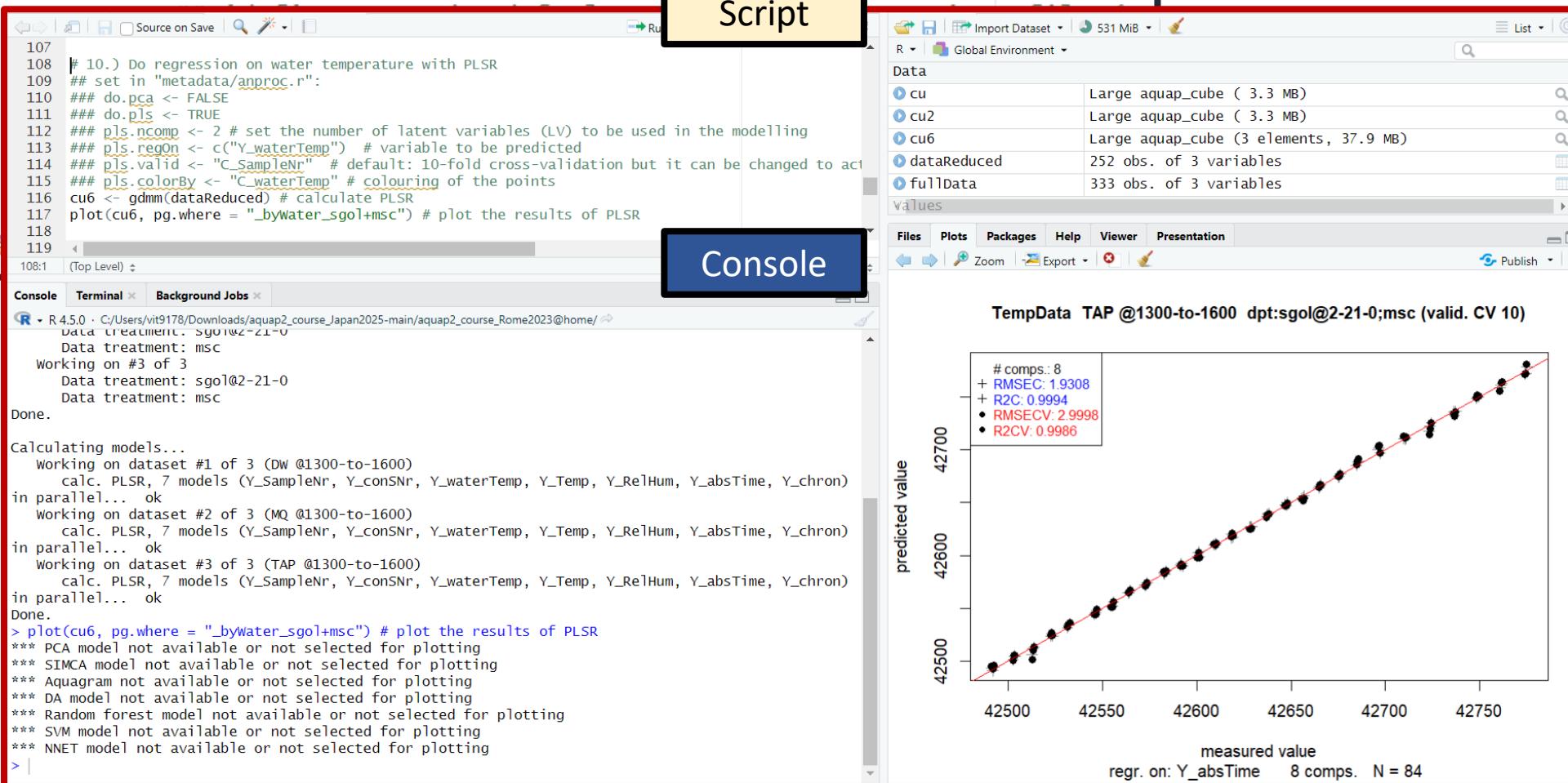
```
35 #####  
36 ## PCA  
37 do.pca <- T ## if PCA of the given datasets  
38  
39 # plotting  
40 pca.colorBy <- NULL  
41 pca.elci <- "def"  
42 pca.elcolorBy <- NULL  
43 pca.what <- "both"  
44 pca.sc <- c(1, 2)  
45 pca.sc.pairs <- 1:6  
46 pca.lo <- 1:6  
47 #####
```



Spectrum-based calculations – PLSR

```
61 #####
62 ### PLSR
63 do.pls <- FALSE
64 pls.ncomp <- NULL
65 pls.region <- NULL
66 pls.valid <- "def"
67 pls.exout <- "def"
68
69 # plotting
70 pls.colorBy <- NULL
71 pls.what <- "both"
72 pls.rdp <- FALSE
73 #####
74
```

```
## if PLSR models of the given datasets should be calculated
## number of components, leave at NULL for automatic detection
## which variables should be used for the regression? Leave empty character vector
```



Spectrum-based calculations – aquagrams

Anproc

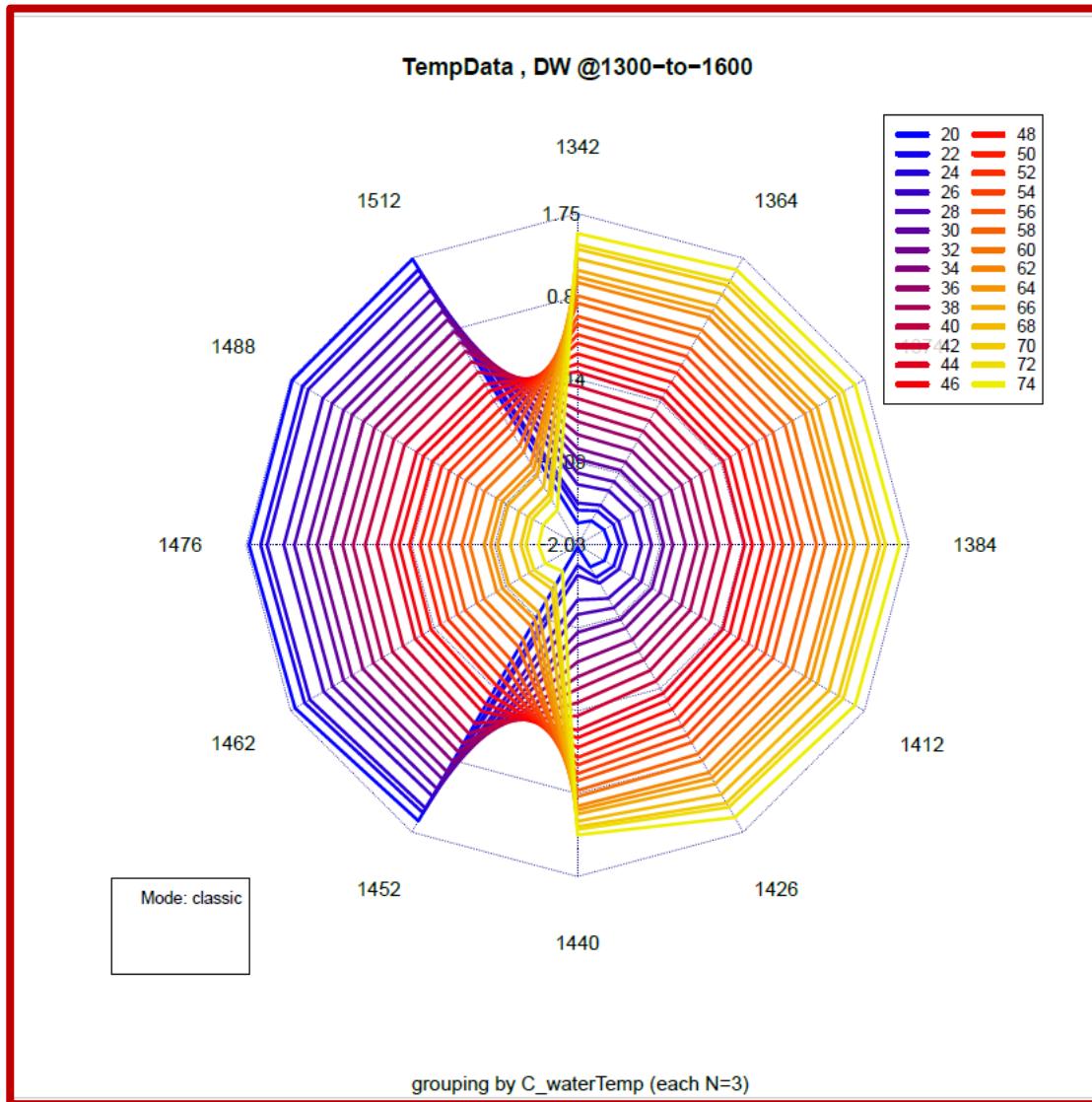
```
77  ### Aquagram
78  do.aqg <- FALSE
79  aqg.vars <- NULL
80  aqg.nrCorr <- "def"
81  aqg.spectra <- FALSE
82  aqg.minus <- NULL
83  aqg.mod <- "def"
84  aqg.TCalib <- "def"
85  aqg.Texp <- "def"
86  aqg.bootCI <- "def"
87  aqg.R <- "def"
88  aqg.smoothN <- 21
89  aqg.selWls <- "def"
90  aqg.msc <- TRUE
91  aqg.reference <- NULL
92
```

Script

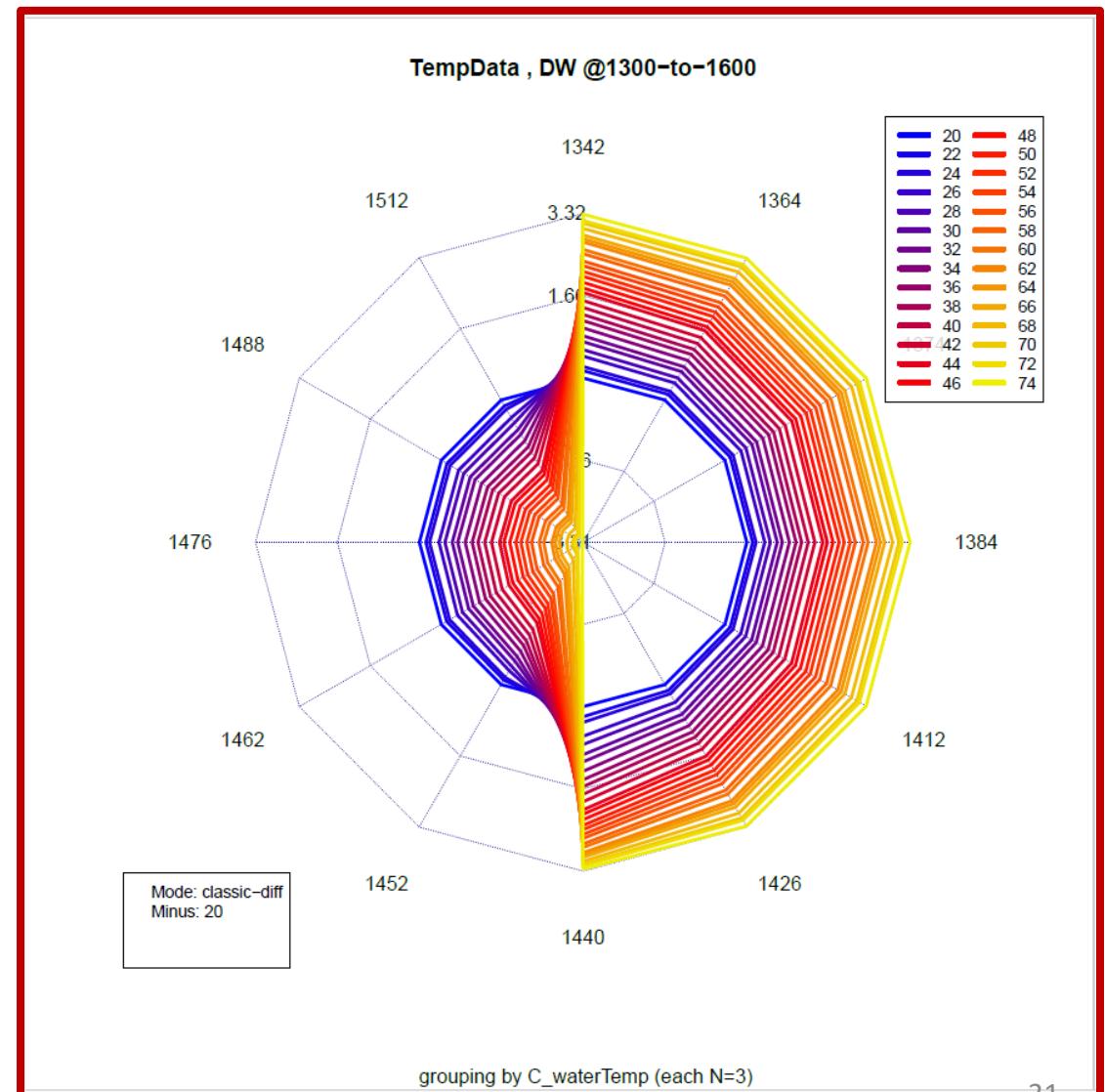
```
119 # 11.) Create classic aquagrams
120 ## set in "metadata/anproc.r":
121 ### do.pls <- FALSE
122 ### do.aqg <- TRUE
123 ### aqg.vars <- "C_waterTemp"
124
125
126
127 # 11a.) Calculate classic aquagram
128 ### aqg.mod <- "classic"
129 cu7 <- gdmm(dataReduced)
130 plot(cu7, pg.fns = "_byWater") # plot "classic" aquagram
131
132
133 # 11b.) set anproc to do classic aquagram by subtracting lowest temperature
134 ## set in "metadata/anproc.r":
135 ### aqg.spectra <- "all"
136 ### aqg.minus <- "20"
137 ### aqg.mod <- "classic-diff"
138 cu8 <- gdmm(dataReduced)
139 plot(cu8, pg.fns = "_byWater") # plot "classic-difference" aquagram
140
```

Spectrum-based calculations – aquagrams

Classic



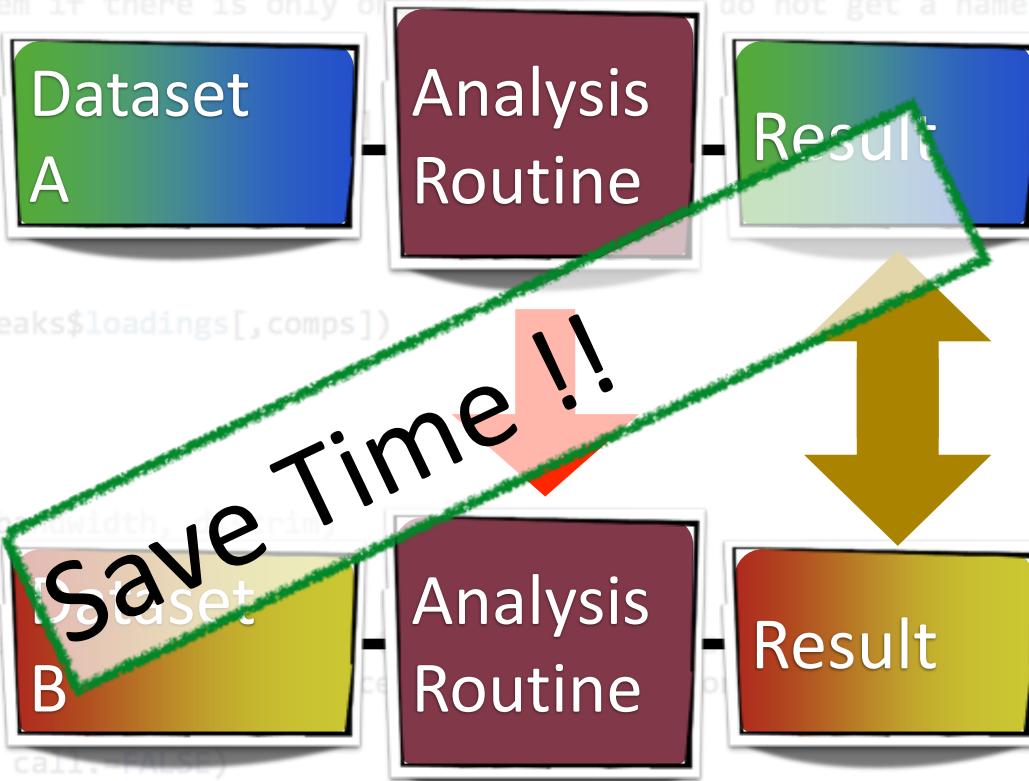
Classic-diff



Summary - Main Advantages

- **fully scriptable** analysis routines
 - consisting of (user's) custom code, and
 - easy-to-use analysis modules

- **same analysis** of an exact repetition of a previous experiment can so be done **in a few seconds** - just by plugging new data into the old (copied) analysis routine



Think - Copy&Paste - Enjoy

Case study: adding data to a sample NIR – R database

Case study

- Individual experiments
 - Presentation of sample experiments
 - Paprika – adulterated with corn flour
 - Powdered sugar – adulterated with wheat flour
 - Grape seed extract – adulterated with plant extracts
 - Quick on spot experiment
 - Paprika – adulterated with corn flour
- Importing the data of individual experiments into a database
- Effective data evaluation



Aquap2 practice

<https://github.com/bpollner/aquap2>

https://github.com/zoltankovacs/NIR2025-HASH_NIR4