

Stepped thermal analysis in accordance with DIN19539

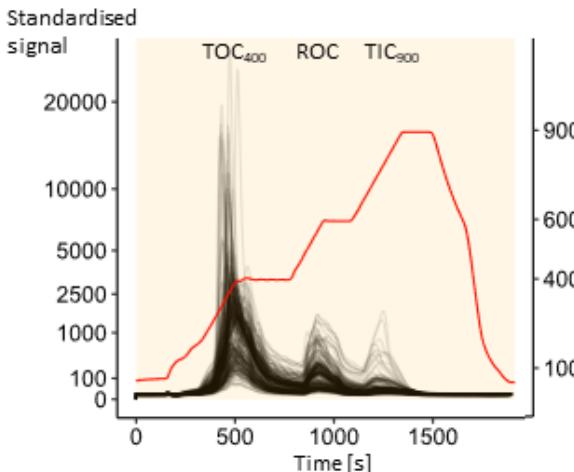
Carbon fractions (DIN19539):

TOC₄₀₀: labile SOC-Pool

ROC: recalcitrant SOC-Pool

TIC₉₀₀: inorganic C-Pool

TOC

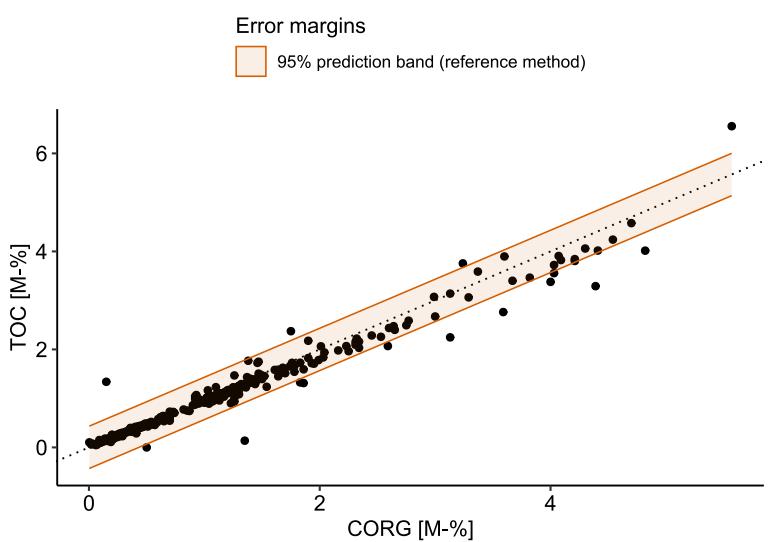


TOC vs CORG

BDF reference data CORG vs soliTOC measurements TOC

Error margins

95% prediction band (reference method)

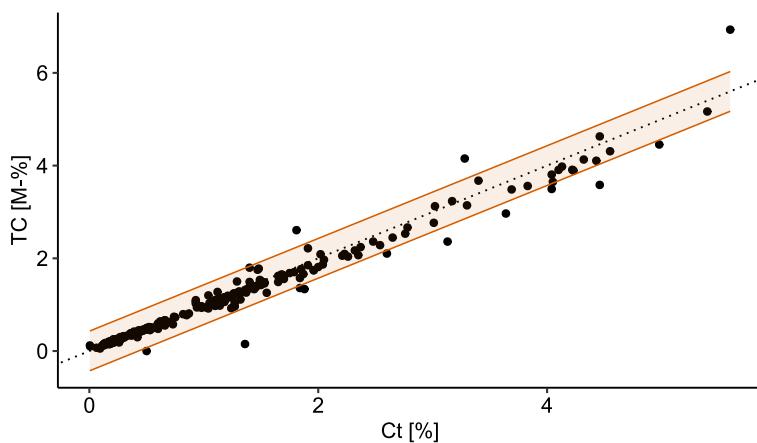


TC vs Ct

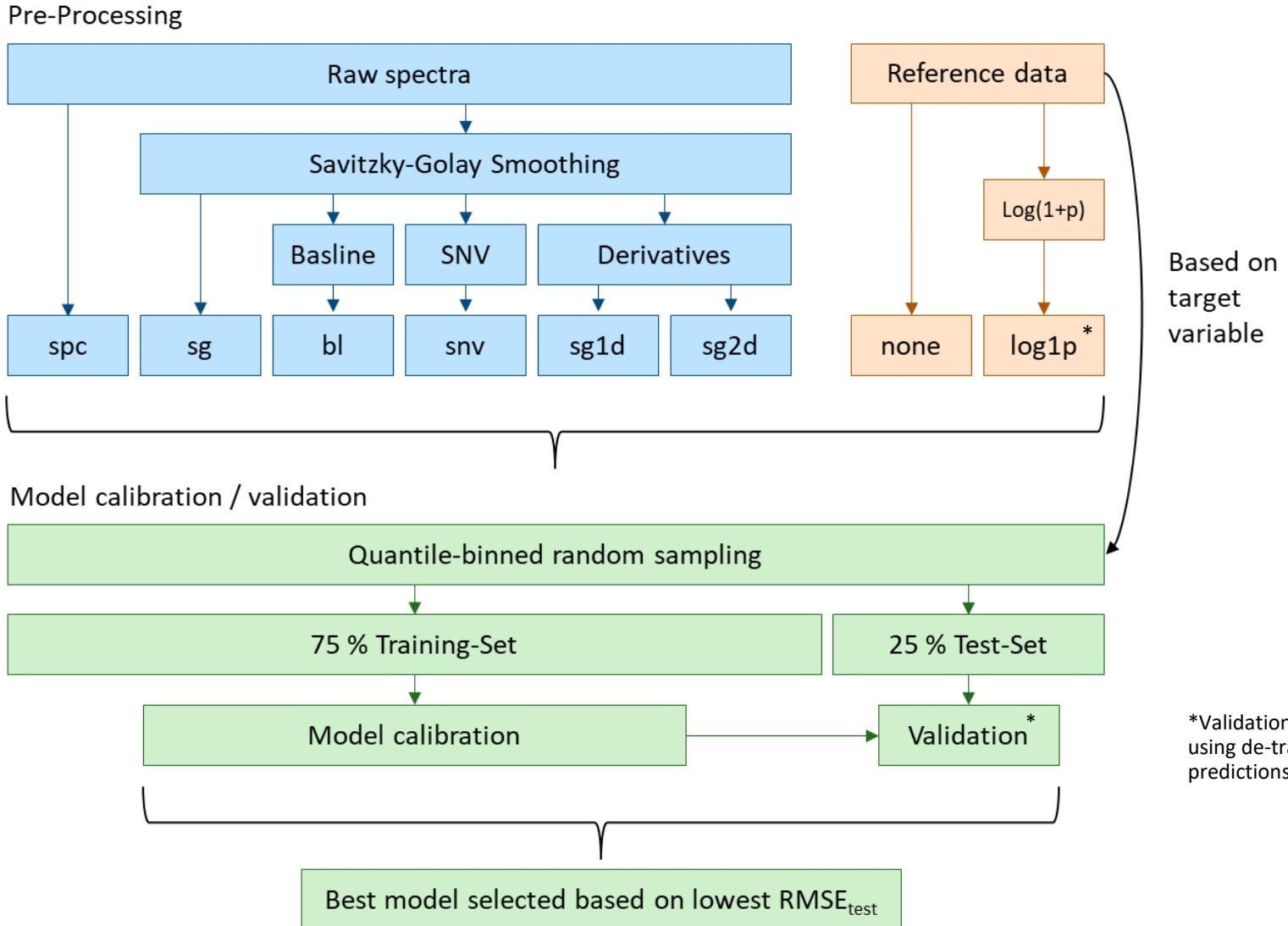
BDF reference data Ct vs soliTOC measurements TC

Error margins

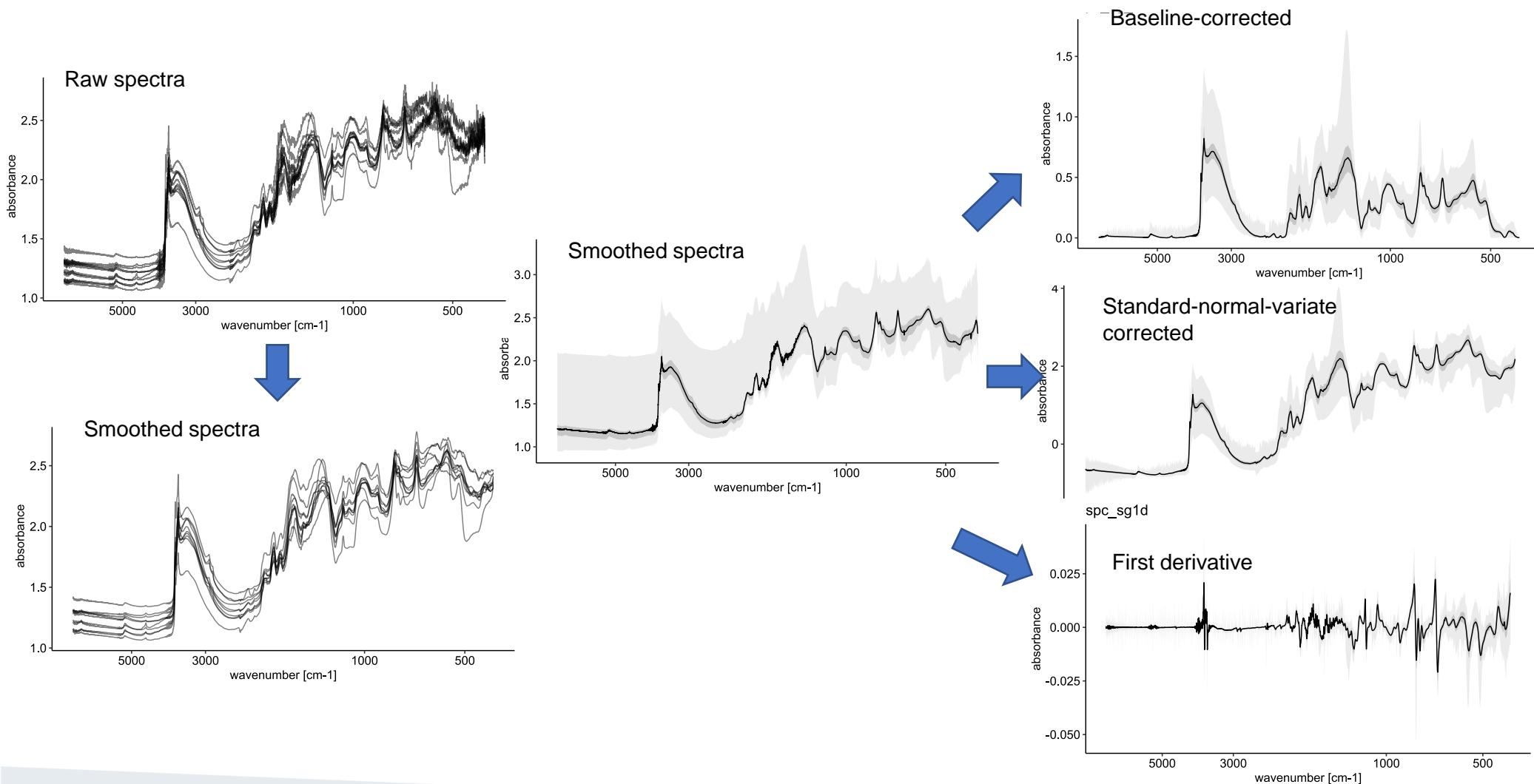
95% prediction band (reference method)



Pre-Processing and model calibration



Visualisation of different pre-processing steps



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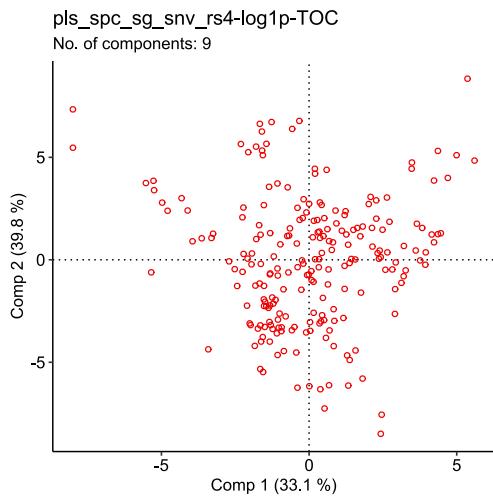
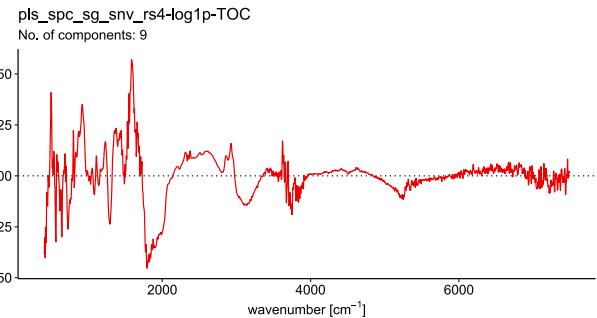
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Sean Adam and Conrad Jackisch Interdisciplinary Environmental Research Centre, TU Bergakademie Freiberg

Different used model types

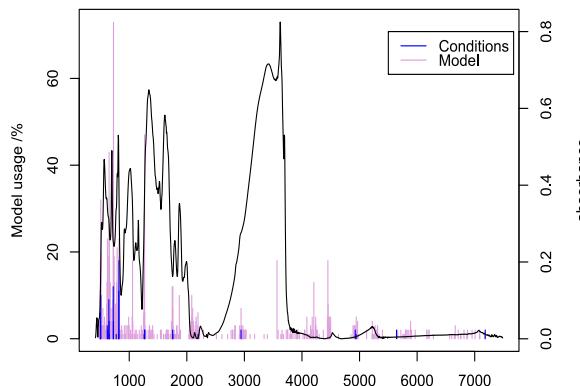
PLS

Partial Least Squares Regression using the NIPALS algorithm



Cubist

Rule-based regression tree with linear models at terminal nodes



```
Model 1:  
Rule 1/1: [3 cases, mean 0.0255620, range 0.0216625 to 0.03139742, est err 0.0059842]  
if  
    3718 > 1.472187  
    1538 <= 1.872462  
then  
    outcome = -0.499281 + 0.6 2014 + 0.37 7354 + 0.42 2026 + 0.26 3190  
    + 0.25 2718 - 0.21 4702 + 0.24 2010 + 0.23 1886 - 0.18 4926  
    + 0.21 1882 - 0.15 3718 + 0.11 4066 + 0.12 2166 - 0.08 1582  
    + 0.05 1230 - 0.06 3762 - 0.05 1254 + 0.04 6686 - 0.05 1614  
    - 0.03 1550 + 0.04 1310 - 0.05 1298 + 0.03 1558 + 0.01 102  
Rule 1/2: [177 cases, mean 0.0848272, range 0 to 0.259892, est err 0.014795]  
if  
    3718 <= 1.472187  
then  
    outcome = -0.4187127 + 14.41 2014 - 3.06 2010 - 7.51 2026 - 3.6 4926  
    + 3.51 1590 - 2.79 4066 - 3.06 1898 + 2.67 1882 + 2.03 4702  
    + 2.03 2166 - 2.03 1230 - 2.03 1558 - 2.03 1298 - 2.03 1550 - 2.03 1558  
    + 1.35 1538 - 1.08 1558 - 1.09 1550 + 0.81 1230 - 0.81 1254  
    - 0.82 2718 + 0.65 1546 - 0.62 1614 - 0.59 1298 + 0.33 6686  
    + 0.43 1310 + 0.14 502 - 0.17 7354 - 0.14 454 - 0.04 2922  
Rule 1/3: [32 cases, mean 0.2535620, range 0.01670498 to 0.608321, est err 0.0434782]  
if  
    4702 <= 1.420133  
    3718 > 1.472187  
    1538 > 1.872462  
then  
    outcome = -0.4187129 + 26.9 2014 - 10.5 2010 - 11.46 2026 + 7.43 1590  
    - 6.53 4926 + 3.67 4702 + 1.76 1538 - 3.64 1583 - 3.35 1546  
    + 3.68 2166 + 2.05 1882 - 1.91 1614 + 1.2 1230 - 1.7 1898  
    - 1.25 1254 + 0.93 1558 + 0.87 7354 - 1.2718 - 0.61 3762
```

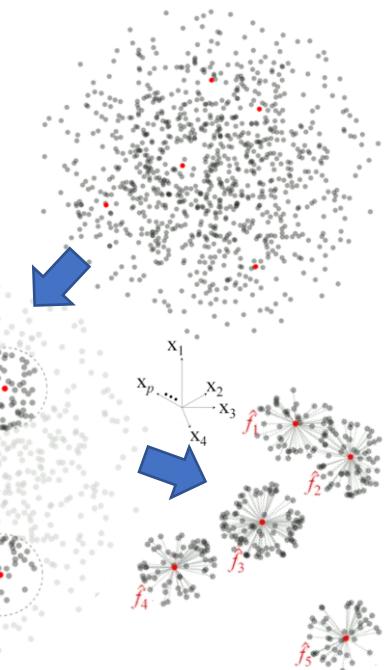
MLB

Memory Based Learner

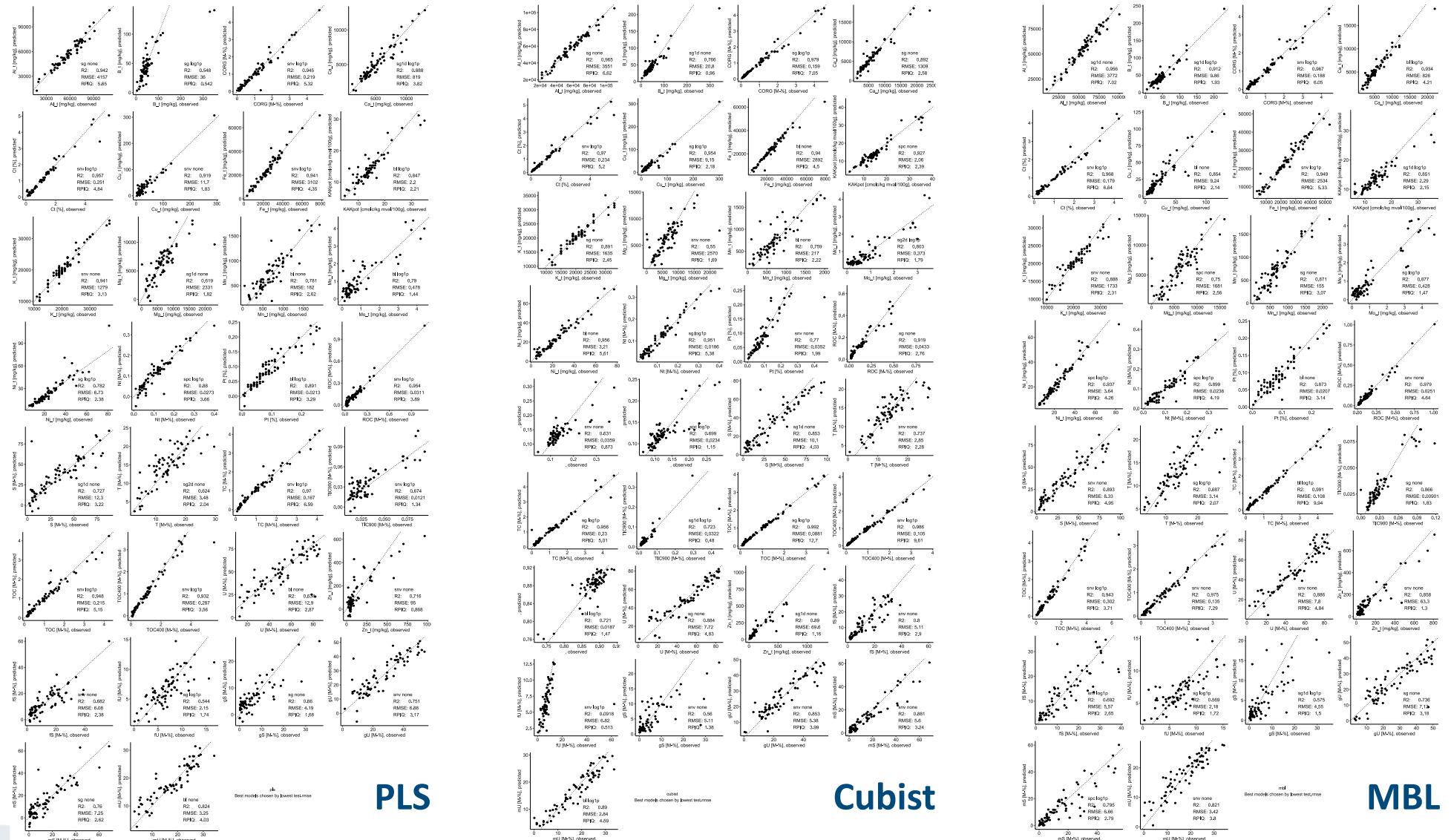
For each predictor spectra, a set of k most similar spectra are selected based on a similarity metric (Mahalanobis distance). For each predictor spectra an individual weighted averaged pls is created

<https://cran.r-project.org/web/packages/ensamble/readme/README.html>
(modified)

● Reference spectra
● Target spectra



Model Validation



PLS

Cubist

MBL

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Soil properties / target variables

Variable (short)	Variable (long)	Unit	Method
TOC ₄₀₀	Labile organic carbon	wt-%	Stepped thermal analysis / DIN19539
ROC	Recalcitrant organic carbon	wt-%	Stepped thermal analysis / DIN19539
TIC ₉₀₀	Inorganic carbon	wt-%	Stepped thermal analysis / DIN19539
TOC	Total organic carbon (TOC ₄₀₀ + ROC)	wt-%	Stepped thermal analysis / DIN19539
TC	Total carbon (TOC+TIC ₉₀₀)	wt-%	Stepped thermal analysis / DIN19539
C _{ORG}	Total organic carbon	wt-%	Determination of total organic carbon
C _t	Total carbon	wt-%	Elemental Analysis (C,H,N,S,O)
N _t	Total nitrogen	wt-%	Elemental Analysis (C,H,N,S,O)
P _t	Total phosphorous	wt-%	EDRFA
gS	Coarse sand	wt-%	Texture distribution
mS	Medium sand	wt-%	Texture distribution
fS	Fine sand	wt-%	Texture distribution
S	Sand	wt-%	Texture distribution
gU	Coarse silt	wt-%	Texture distribution
mU	Medium silt	wt-%	Texture distribution
fU	Fine silt	wt-%	Texture distribution
U	Silt	wt-%	Texture distribution
T	Clay	wt-%	Texture distribution
KAK _{pot} (CEC _{pot})	Cation exchange capacity	cmol _e kg ⁻¹	BaCl ₂ , exchangeable cations DIN196848
Al _t	Total aluminium	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
B _t	Total boron	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
Ca _t	Total calcium	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
Cu _t	Total copper	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
Fe _t	Total iron	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
K _t	Total potassium	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
Mg _t	Total magnesium	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
Mn _t	Total manganese	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
Mo _t	Total molybdenum	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
Ni _t	Total nickel	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations
Zn _t	Total zinc	mg kg ⁻¹	HF/HNO ₃ /HCl microwave digestion, determination of total elemental concentrations

Re-analysis using the Elementar soliToc

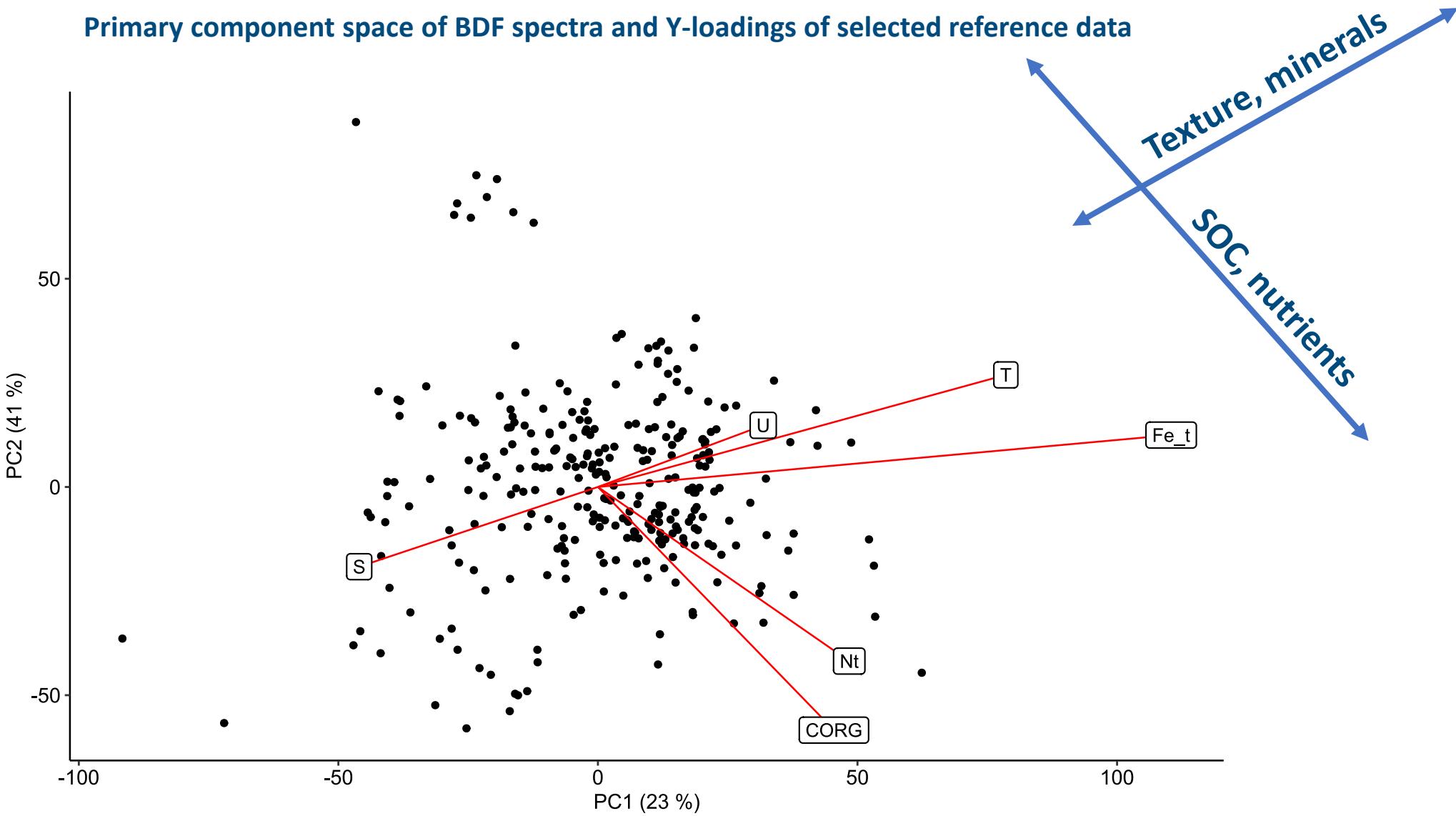
Reference values obtained from the BDF database.

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Primary component space of BDF spectra and Y-loadings of selected reference data



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