



Predictive Modeling of Soil Properties (Total Nitrogen and pH) Using MIR Spectroscopy.

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BACKGROUND

Mid-infrared (MIR) spectroscopy has emerged as a non-destructive, eco-friendly, and cost-efficient tool for soil analysis, offering high resolution and wide absorbance range capabilities. Integrating MIR data with machine learning, especially Partial Least Square Regression (PLSR), has shown excellent predictive capabilities for various soil properties such Total Nitrogen (TN) and pH. However, the accuracy of these predictions can be affected by factors like moisture, soil particle size, and baseline drift, necessitating effective data preprocessing.

AIM OF THE PROJECT

This study evaluates the impact of various preprocessing techniques and variable selection methods, such as SNV and Savitzky-Golay filters in combination with variable selection techniques, on improving the prediction accuracy of Total Nitrogen and pH.

MATERIAL AND METHODS

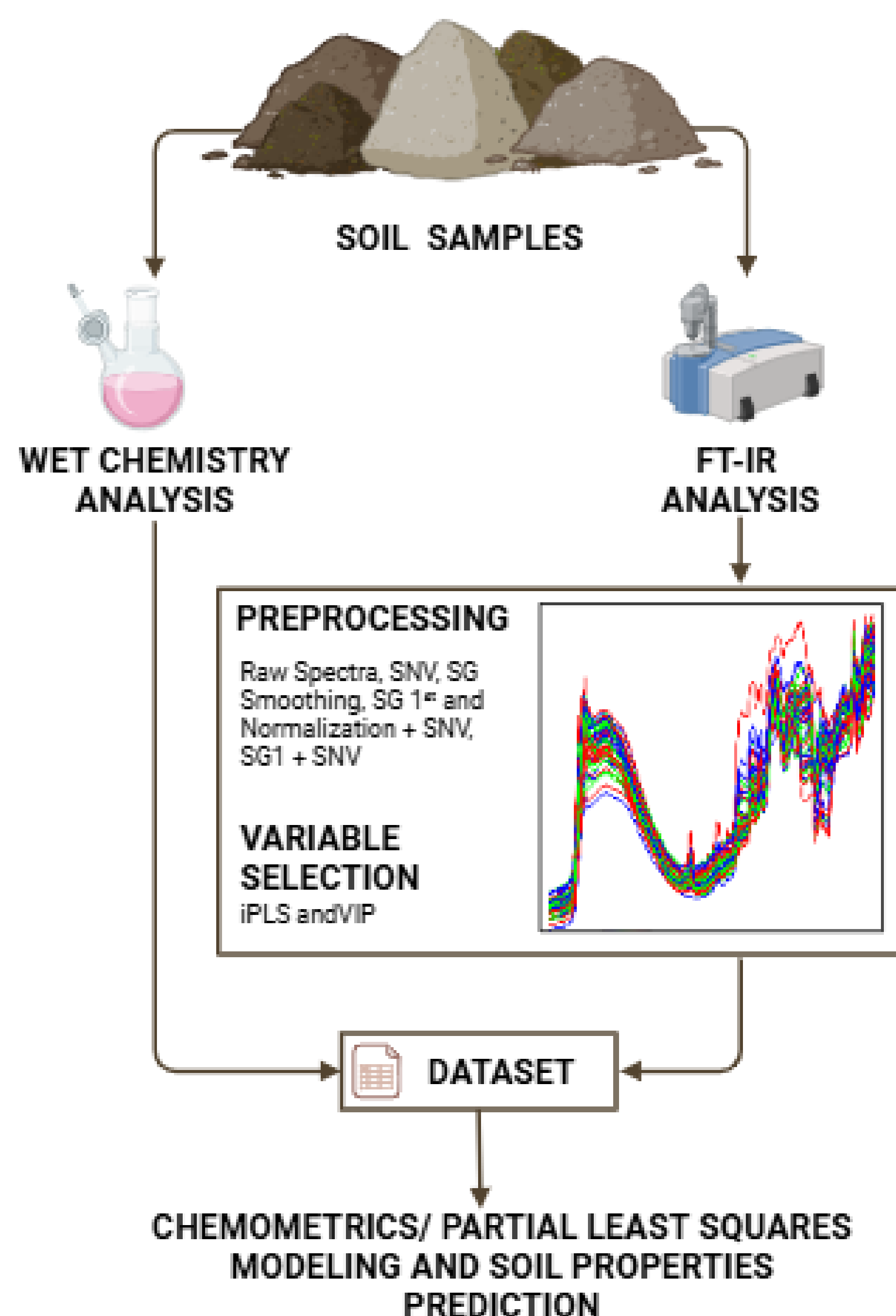


Fig.1: Workflow for evaluating the impact of preprocessing and its combination with variable selection on PLSR models for estimating TN and pH.

RESULTS

FTIR SPECTRA OF SOILS

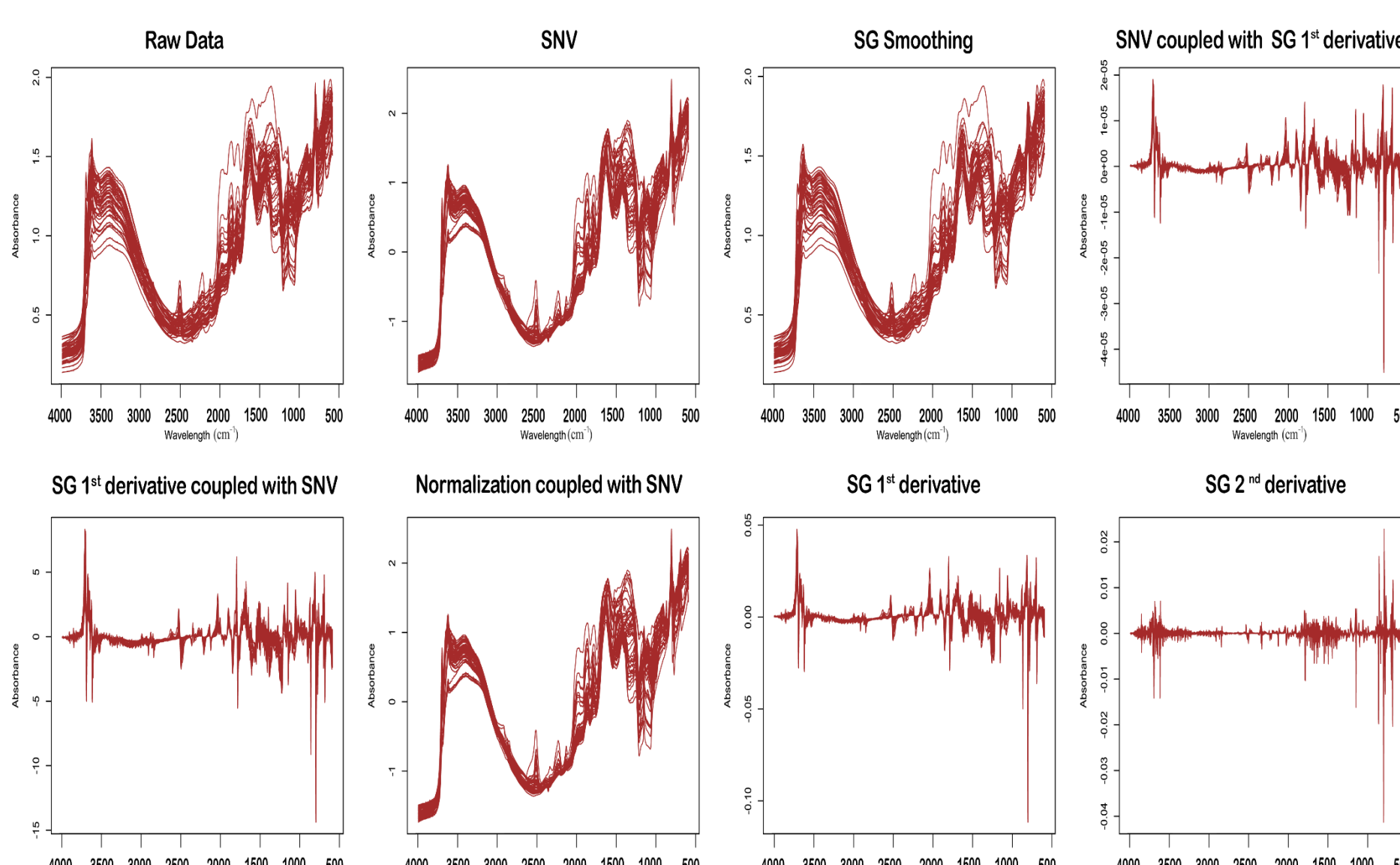


Fig. 2: FTIR spectra of the soil samples before preprocessing and after applying different statistical pretreatment method

SOIL PROPERTIES PREDICTION:

TN and pH were predicted using raw data and data subjected to various preprocessing techniques to identify the optimal preprocessing method for improving prediction accuracy. Then, the best preprocessing method was selected to apply the variable selection methods (VIP and iPLS).

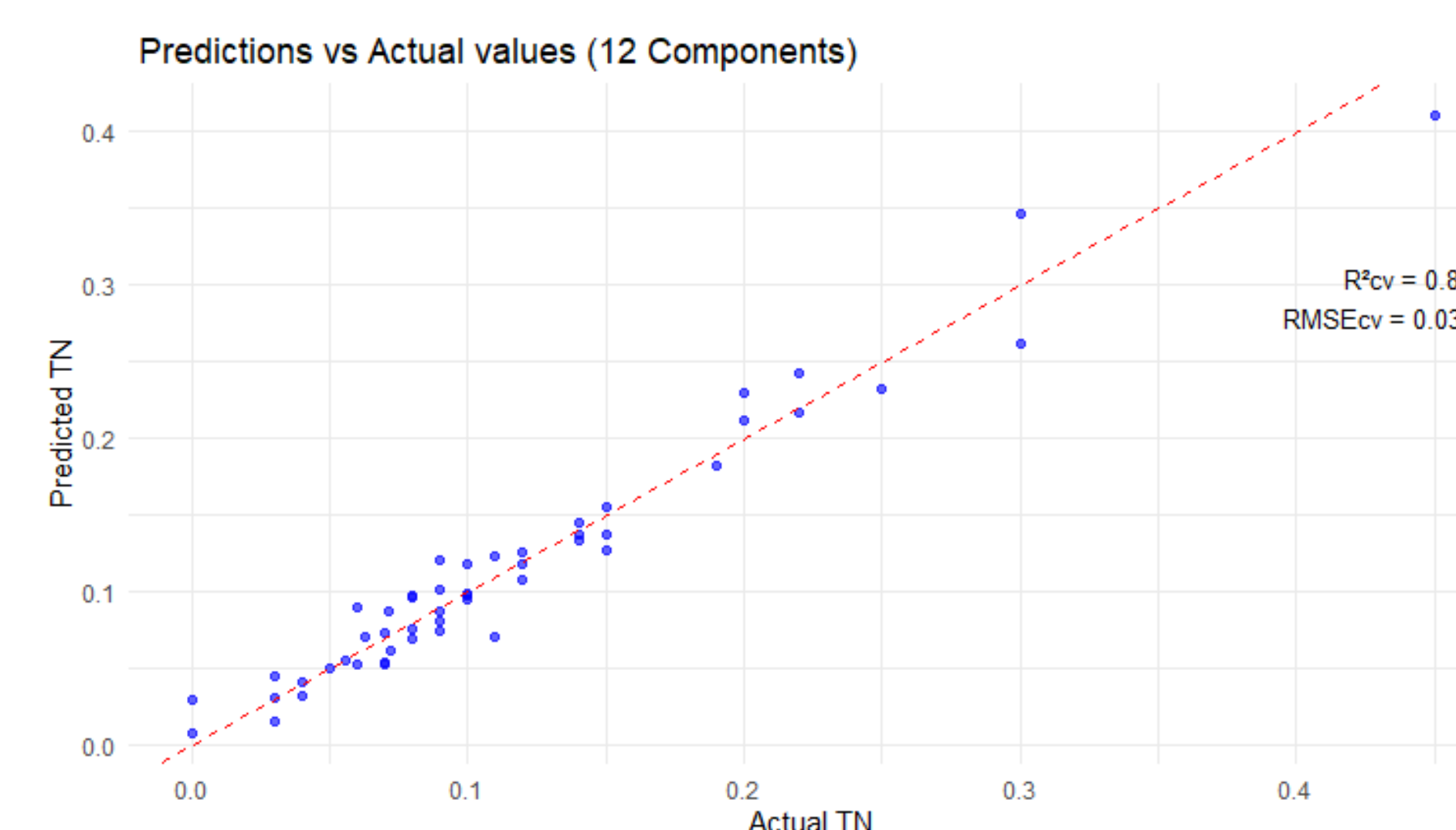


Fig. 3: The prediction of TN using Raw data (Unprocessed data)

Table 1: Table of merit of the prediction of TN and pH using MIR data and preprocessing data with the application of iPLS and VIP.

| | Raw Data | | SNV | | SG Smoothing | | NOR+SNV | | SG1 | |
|----|----------|--------|----------|--------|--------------|--------|---------|--------|------|--------|
| | R2 | RMSEcv | R2 | RMSEcv | R2 | RMSEcv | R2 | RMSEcv | R2 | RMSEcv |
| TN | 0.82 | 0.03 | 0.8 | 0.03 | 0.81 | 0.03 | 0.83 | 0.03 | 0.87 | 0.01 |
| pH | 0.66 | 0.36 | 0.69 | 0.35 | 0.61 | 0.39 | 0.73 | 0.2 | 0.71 | 0.33 |
| | SG1+SNV | | SG1+iPLS | | SG1+VIP | | | | | |
| | R2 | RMSEcv | R2 | RMSEcv | NI | TRC | R2 | RMSEcv | TRC | |
| TN | 0.83 | 0.03 | 0.92 | 0.02 | 8 | 2.5 | 0.91 | 0.02 | | |
| pH | 0.66 | 0.37 | 0.93 | 0.15 | 73 | 3 | 0.86 | 0.23 | | |

TN: Total Nitrogen,, SNV: standard normal variate, NOR: Normalization, SG S: Savitzky–GolaySmoothing, SG1: Savitzky–Golay 1st Derivative, VIP: Variable Importance in Projection, iPLS: Intervals Partial least squares.

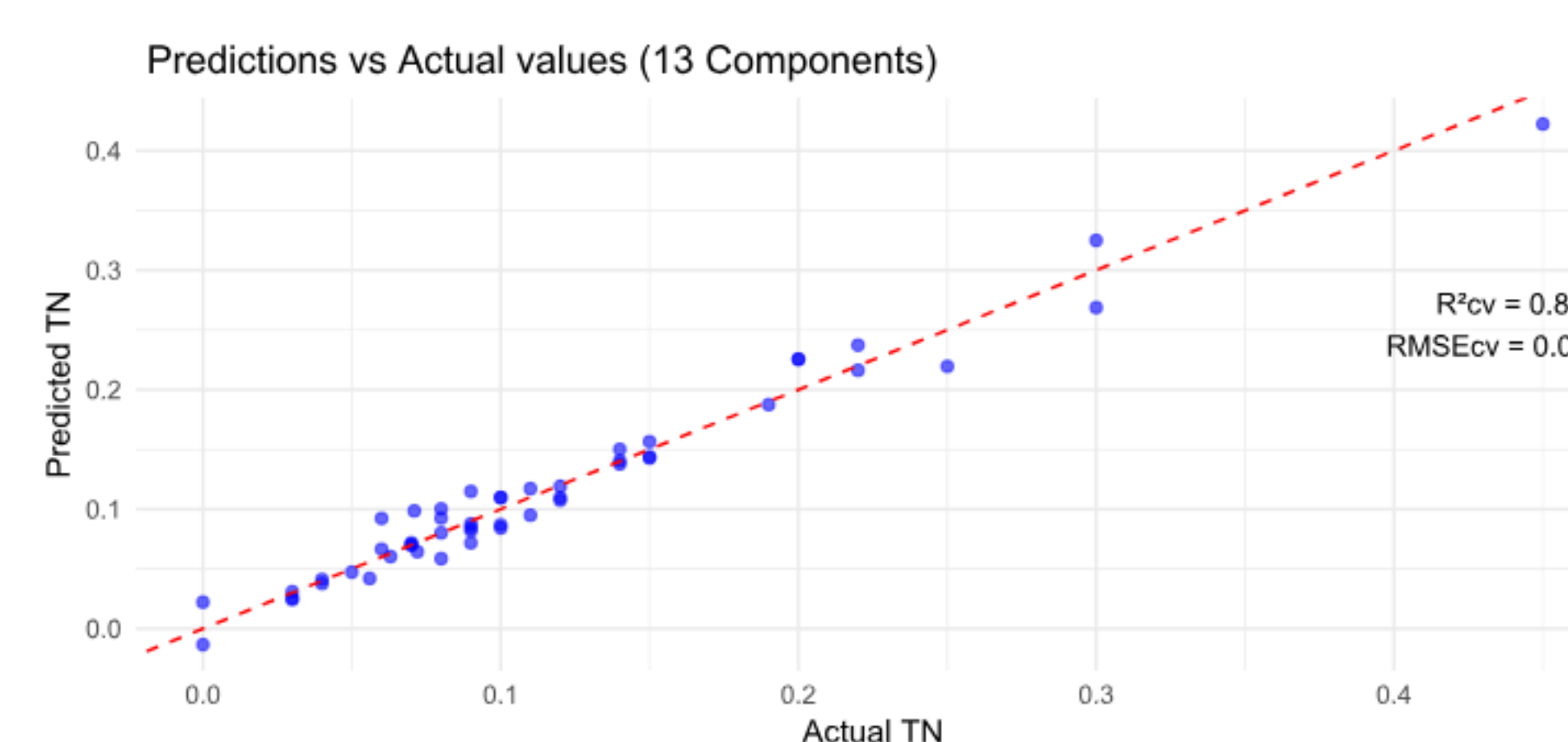


Fig. 4: The prediction of TN using Savitzky-Golay 1st derivative

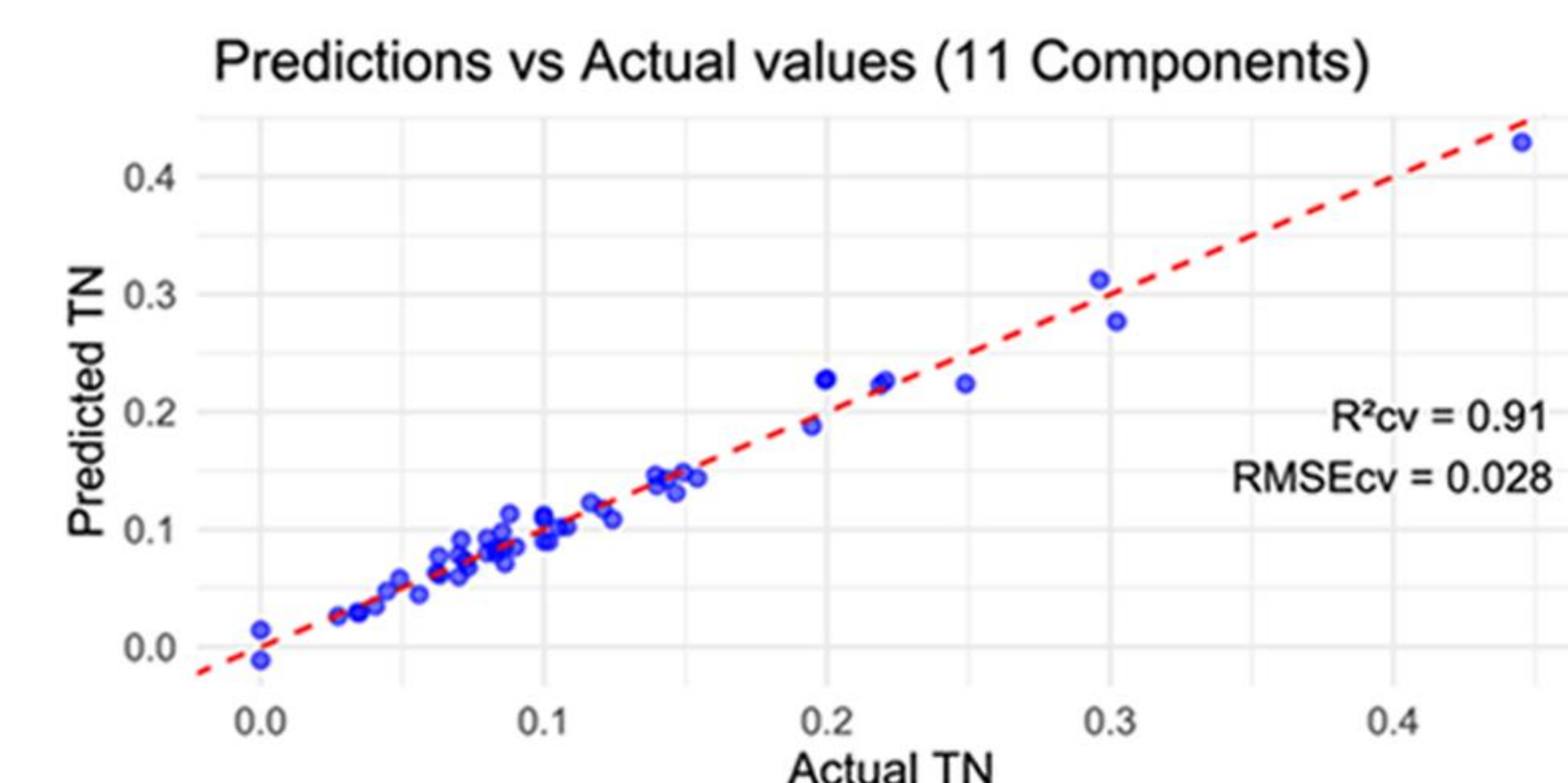


Fig. 5: The prediction of TN using Savitzky-Golay 1st derivative Combined with variable in proection (VIP)

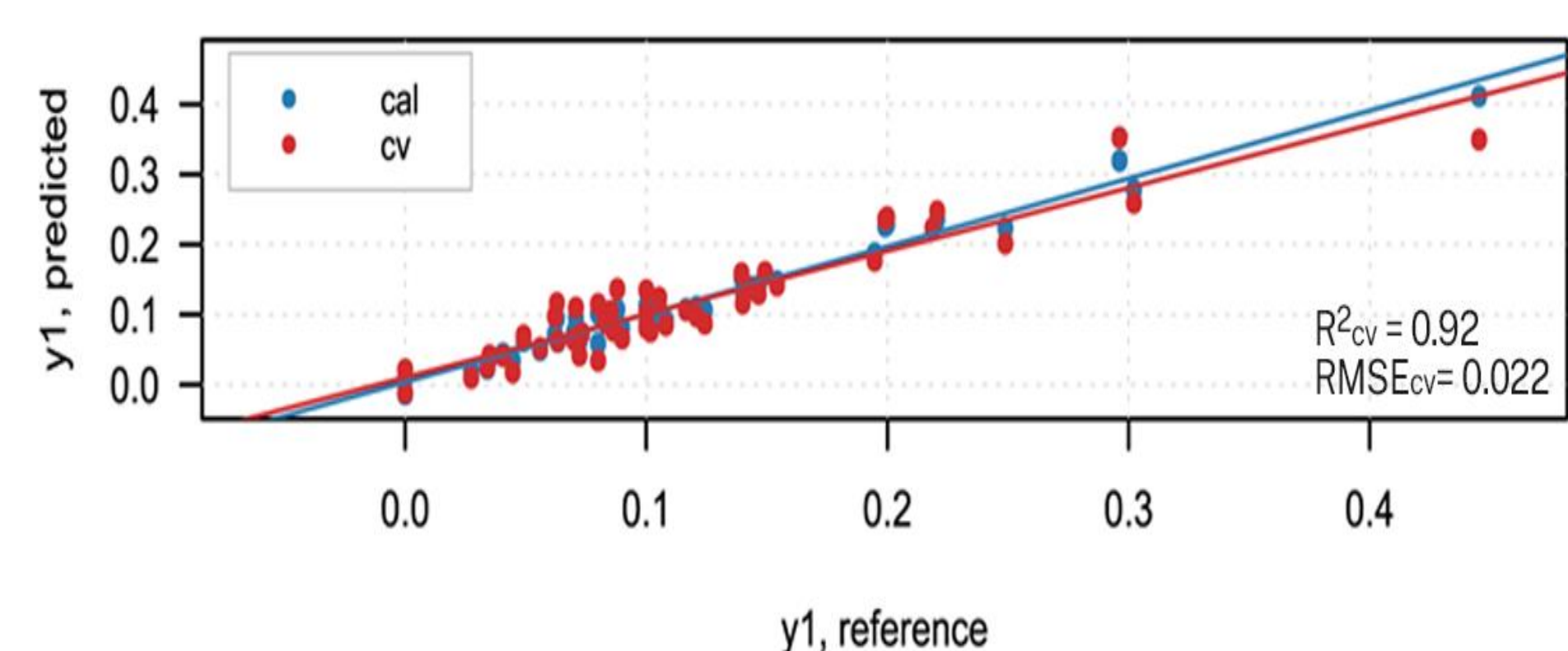


Fig. 6: The prediction of TN using Savitzky-Golay 1st derivative and interval-PLS (iPLS).

CONCLUSION

The Savitzky-Golay first derivative has proven to be the most efficient preprocessing method for improving the prediction of TN and pH using a Partial Least Squares Regression (PLSR) model. Due to these findings, SG1 was chosen as the optimal preprocessing method to be combined with VIP and iPLS approaches for variable selection. iPLS demonstrated excellent improvements for TN and pH, while VIP also yielded excellent results but iPLS performed better. Combining variable selection methods with adequate preprocessing significantly improved the prediction of the studied soil properties.

Acknowledgement

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