

# Estimating the mineralizable form of organic carbon (OC) in bio-based fertilizers using non-destructive spectroscopy techniques

Khan Wali\*, Haris Ahmad Khan\*, Mark Farrell\*\*, Eldert van Henten\*, Erik Meers\*\*\*

- Farm Technology group, Wageningen University & Research, the Netherlands
- \*\* CSIRO Agriculture and Food, Karna Country, Locked Bag 2, Glen Osmond, SA 5064, Australia
- \*\*\* Department of Green Chemistry and Technology, University of Gent, Belgium

## Abstract

The estimation of organic carbon (OC) mineralization is desirable for both agronomic and environmental reasons. To estimate the rate of mineralization, the NIR and MIR spectroscopy has been used in past, which link the rate of mineralization with the composition of biobased fertilizers. The current study evaluate the potential of NIR and MIR spectroscopy in combination with machine learning techniques to estimate the mineralizable form of OC to  $CO_2$ . Carbon mineralization from 85 different potential soil organic amendments (composts, manures, plant residues and biosolids) was quantified under controlled environmental conditions over a 547-day incubation period. Cumulative mineralization of organic carbon from the amendments was fitted to a two-pool exponential model. A wavelength selection method followed by partial least squares regression was adapted to estimate the fast-decomposing pool ( $C_f$ ) and slow decomposing pool ( $C_s$ ). The  $C_f$  and  $C_s$  values has been estimated with coefficient of determination value ( $R^2=0.94$ ) and root mean square error ( $RMSE=19.17$ ). The obtained result suggests that NIR and MIR spectroscopy has the potential estimate and quantify the plant available for form of diverse set of biobased fertilizers.

## Objectives

- To Quantify the mineralization form of OC to  $CO_2$  using NIR and MIR spectroscopy
- Application advanced machine learning techniques to improve the estimation

## Sensor Technologies

- Visible Near Infrared (Vis-NIR)
- Mid Infrared (MIR)

## Material and Method

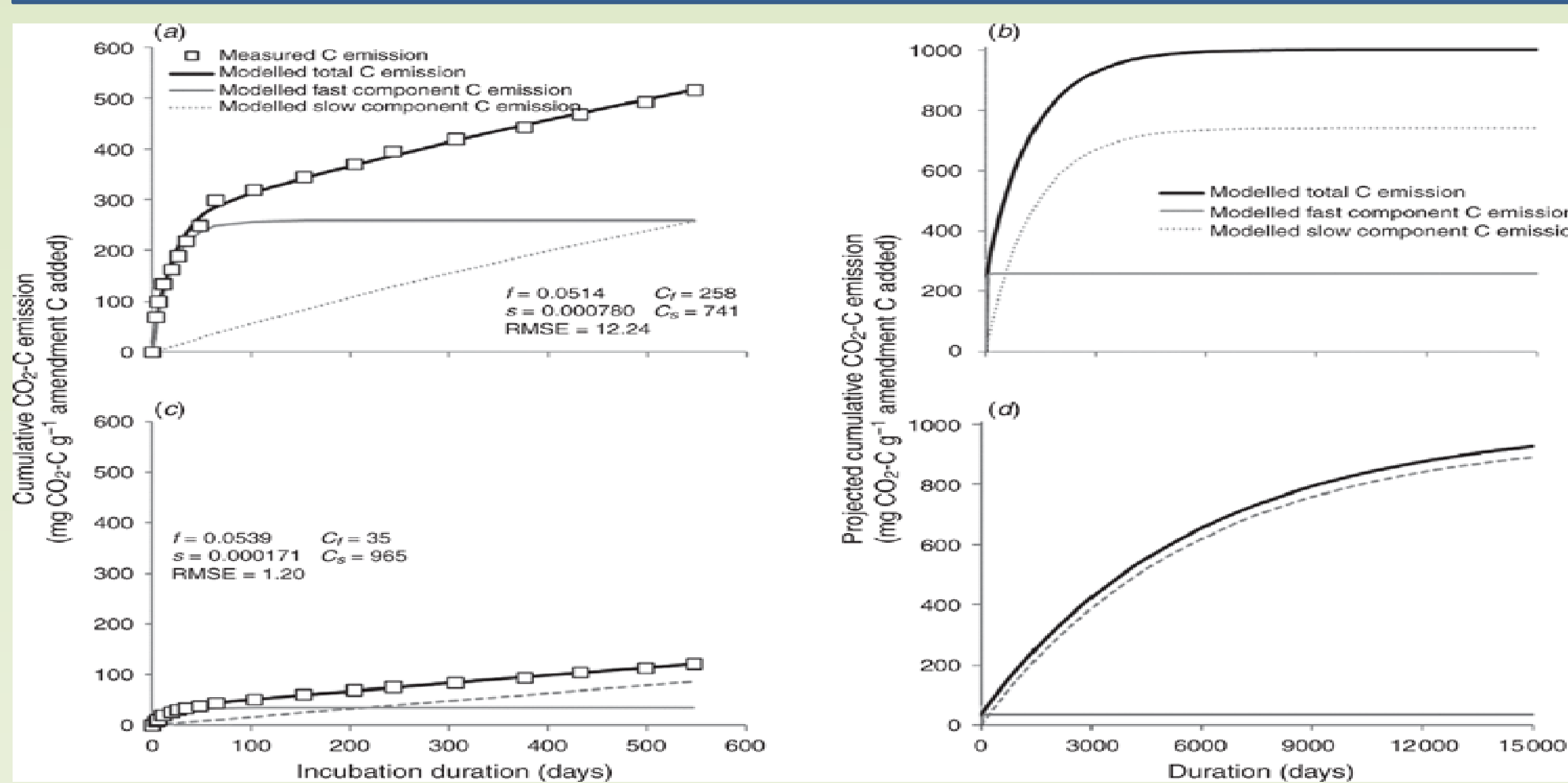


Fig 1. Measured and modelled cumulative  $CO_2$ -C emission from organic amendments exhibiting high (a) and low (c)  $CO_2$ -C emission over the duration of the incubation. Projected cumulative mineralisation beyond the incubation duration derived for the same high (b) and low (d)  $CO_2$ -C emitting amendments.

$$C_{min}(t) = C_f(1 - e^{-ft}) + C_s(1 - e^{-st}) \quad (1)$$

When  $t=0$ ,  $C_{min}(0) = 0$ , and when  $t=\infty$  equation 1 become:

$$C_{min}(\infty) = C_f(1 - 0) + C_s(1 - 0) = C_f + C_s \quad (2)$$

Hence summation of  $C_f$  and  $C_s$  gives total mineralized OC  
Where  $C_s = 1 - C_f$

Fitted two-pool first order exponential decay model

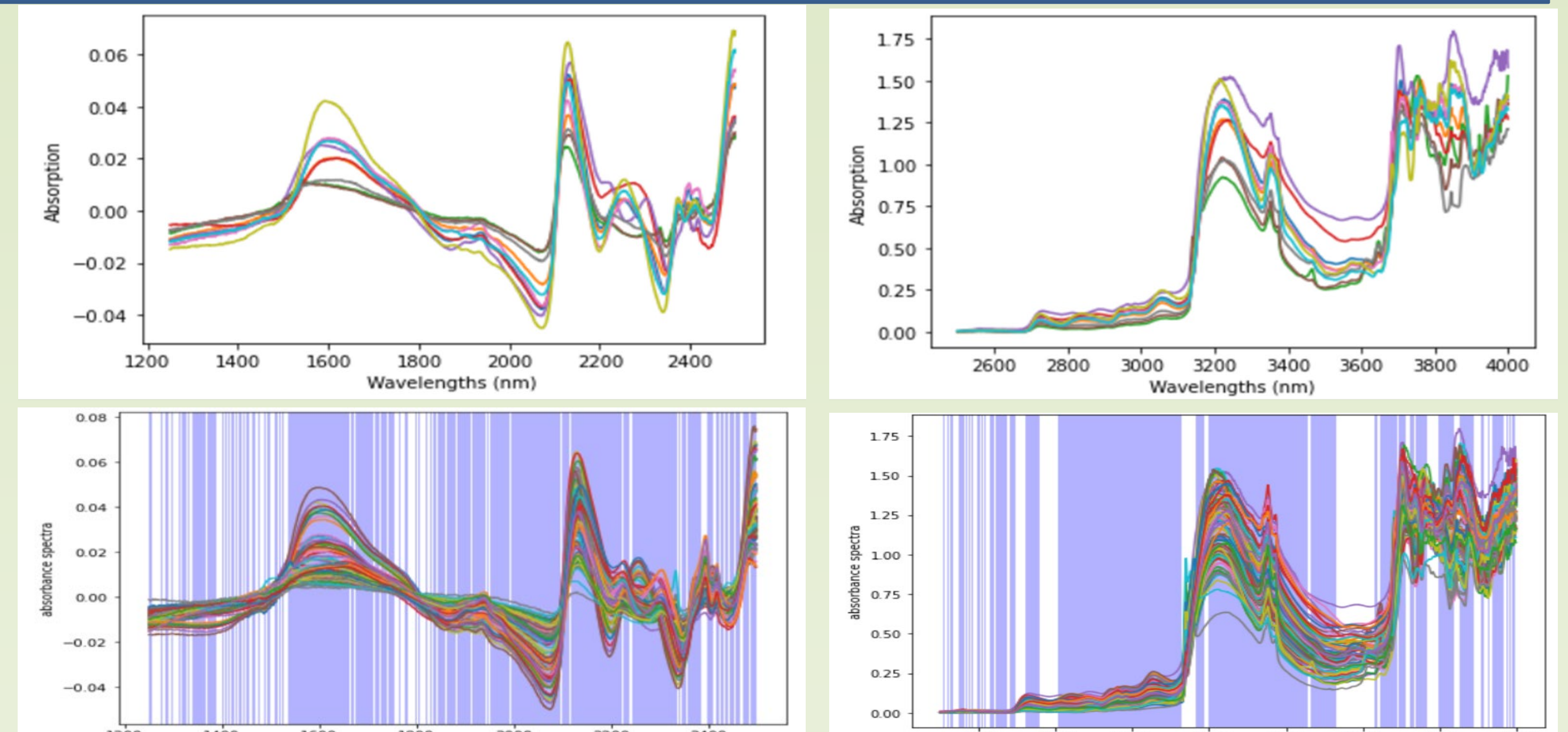


Fig 2. NIR spectral data

Fig 3. MIR spectral data

```

a=1
W(:,a)=X'*Y;
W(:,a)=W(:,a)/norm(W(:,a));
T(:,a)=X*W(:,a);
P(:,a)=X'*T(:,a)/(T(:,a)'*T(:,a));
Q(a,1)=Y'*T(:,a)/(T(:,a)'*T(:,a));
X=X-T(:,a)*P(:,a)';
Y=Y-T(:,a)*Q(a,1);
W=W*(P'*W)^(-1);
a=a+1;
B=W*Q;
Y=Y+X*B;
    
```

```

Algorithm 1 Wavelength selection algorithm
H
1: Fit PLS model on NIR/MIR spectral data and optimize the number latent variables
2: Find all the regression coefficients(B)
3: Arrange B in descending order
4: Select  $\lambda_i$  corresponding to first entry of B i.e  $b_i$ 
5: Fit PLS and find RMSE -> choose this as a reference
6: procedure WAVELENGTH SELECTION( $\lambda_i$ )
7: Fit PLS on  $b_i(k)$  corresponding wavelength where  $k = 2$ 
8: if  $RMSE(k) \leq RMSE(k-1)$  then
9:   Select wavelength
10:   $k=k+1$ 
11:  Repeat step 7
12: else
13:   if  $RMSE(k) > RMSE(k-1)$  then
14:     Discard wavelength
15:     Stop
16:     Print all the selected wavelengths
    
```

Algorithm for wavelength selection and PLS

## Results

Statistic	NIR Spectra				MIR Spectra			
	$C_f$	f	$C_s$	s	$C_f$	f	$C_s$	s
$R^2$	0.94	0.42	0.94	0.15	0.92	0.37	0.91	0.11
RMSE	19.17	0.018	19.17	0.000142	23.56	0.026	23.56	0.00024

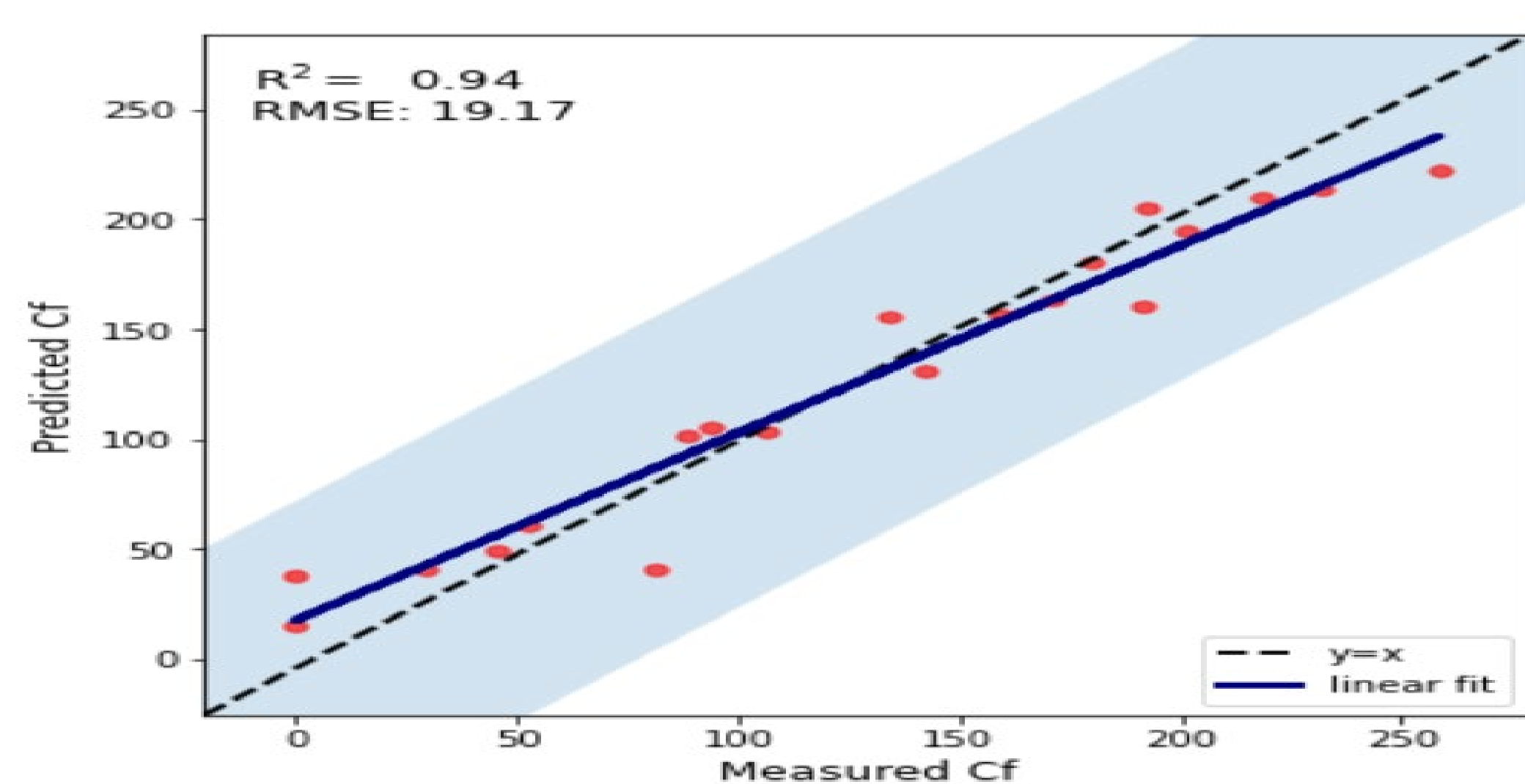


Fig 4. predicted vs actual fast-decomposing pool ( $C_f$ )

## References

1. J. Baldock, C. Creamer, S. Szarvas, J. McGowan, T. Carter, and M. Farrell, "Linking decomposition rates of soil organic amendments to their chemical composition," Soil Research, 2021.
2. Patrick Kiiti Mutuo, Keith D. Shepherd, Alain Albrecht, Georg Cadisch, "Prediction of carbon mineralization rates from different soil physical fractions using diffuse reflectance spectroscopy", Soil Biology and Biochemistry, 2006.

## Conclusion

The potential of near infrared and mid-infrared spectroscopy for the estimation of the rate of OC mineralization to  $CO_2$  from biowaste has been evaluated in the current study. Based on our obtained results, the following conclusions are made.

- 1) NIR and MIR Spectroscopy have the potential to predict rate of mineralization.
- 2) NIR and MIR in combination with machine learning can be a useful tool to predict the mineralizable form of OC.
- 3) In future more experiment will be performed to quantify the amount of nitrogen which will be eventually transformed to plant available form.

The results obtained in this study are beneficial for evaluating both the agronomical as well environmental effects of bio-based fertilizers.

## Acknowledgement



"This project has received funding from the European Union's Horizon 2020 research and innovation programme under the Marie Skłodowska-Curie grand agreement No.860127"