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# Efficient Quantification of Soluble and Insoluble Oxalates in Clay Mineral Mixtures

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#### ABSTRACT

Quantification of oxalate salts in soil clay minerals is necessary to study oxalate biogeochemistry, but existing analytical techniques are expensive and time-consuming. We aim to develop an efficient attenuated total reflectance mid-infrared (MIR) spectroscopic technique to quantify oxalate salts in a clay mineral matrix. We calibrated MIR models for analysis of oxalate anion concentrations in standard solutions (0-0.01 M) by using a partial leastsquares regression algorithm. MIR models were also developed for analysis of sodium oxalate (NaOx) and calcium oxalate (CaOx) content in clay mineral mixtures with composition like soils of a semi-arid region and with contrasting concentrations  $(0-1.0 \text{ g g}^{-1}$  for both oxalate salts) to test sensitivity of analyses. Validation plots (true vs predicted values) showed excellent model fit ( $R^2 \ge 0.96$ ) and accuracy (normalized root mean squared error of prediction  $\leq$  0.06) for CaOx, NaOx and oxalic acid components. Once predictive models are stored in analytical software, MIR spectroscopic analyses of samples are much more efficient than chemical techniques. Our MIR spectralbased models are suitable for direct quantification of oxalate salts in clay mineral mixtures for samples like those used for model calibration.

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#### **KEYWORDS**

Insoluble oxalates; nondestructive analysis; soluble oxalates; whewellite

# Introduction

Oxalate salts are the most abundant ionic biominerals on Earth (Cheng et al. 2016; Dauer and Perakis 2013). The stability of the calcium oxalate mineral on geological timescales as well as their biological origin make these minerals a useful tracer of geochemical transformations of organic matter (Cheng et al. 2016; Hofmann and Bernasconi 1998). We use the term Ca oxalate or 'CaOx' in the same sense as Uren (Uren 2018), which includes forms in both hydration states, whewellite (CaC<sub>2</sub>O<sub>4</sub>·H<sub>2</sub>O) and weddellite (CaC<sub>2</sub>O<sub>4</sub>·2H<sub>2</sub>O). Calcium oxalates are synthesized by nearly all plants (Franceschi and Horner 1980a) and fulfill various structural and biological functions (Arnott and Webb 2013; Modenesi et al. 2000; Nakata 2003; Pylro et al. 2013; Ross et al. 1999; Tooulakou et al. 2016). CaOx may influence short- and long-term carbon (C) and nutrient supply in ecosystems due to deposition of leaf litter in soils (Dauer and Perakis 2014) and oxalic acid exudation by plant roots (Graustein, Kermit, and Sollins 2011; Jilling et al. 2021; Johnston and Vestal 1993; Pantigoso et al. 2021; Vives-Peris et al. 2020). The role of CaOx in ecosystem calcium (Ca) and C cycling is receiving more attention due to a growing awareness

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of the importance of biomineralization as a soil carbon sequestration mechanism (Bailey, Buso, and Likens 2003; Braissant et al. 2004; Cailleau, Braissant, and Verrecchia 2011a, 2014; Dauer and Perakis 2014; Hervé et al. 2021; Krieger et al. 2017; Martin et al. 2012; Rehman et al. 2021; Uren 2018). Calcium carbonate (CaCO<sub>3</sub>) formation is favored by degradation of CaOx by oxalotrophic bacteria accompanied by an increase in soil pH in a process known as the oxalate-carbonate pathway (OCP) (Braissant, Verrecchia, and Aragno 2002; Cailleau et al. 2014; Uren 2018; Verrecchia, Braissant, and Cailleau 2006) which may be an important mode of atmospheric carbon sequestration (Cailleau, Braissant, and Verrecchia 2011b; Francis and Poch 2019; Hervé et al. 2021; Pons et al. 2018; Rowley et al. 2017; Uren 2018).

The fate of CaOx in soil depends on factors such as soil pH, clay sorption properties, interactions with redox reactive compounds and microbial activity (Dutton and Evans 1996). However, turnover rates of CaOx in soils remain largely unknown (Dauer and Perakis 2013, 2014), as accurate quantification of CaOx has proven to be a significant obstacle. Oxalates in plants occur mainly in the form of insoluble CaOx crystals ( $K_{sp}$  2.57 × 10<sup>-9</sup>) as well as soluble oxalic acid and sodium or potassium oxalate salts (with sodium oxalate or 'NaOx' being the most common water-soluble form in plants) (Franceschi and Horner 1980b; Ilarslan et al. 1997; Liu et al. 2015). Distinction between insoluble and soluble oxalates is thus crucial for oxalate quantification procedures in studies of OCP function, which operates via transformation of insoluble CaOx. A high-throughput, cost-effective method for measuring oxalate salts in geological materials would advance our understanding of the importance of CaOx in environmental carbon and nutrient cycles.

Current techniques to quantify oxalates include Raman spectroscopy and analysis of sample extracts by colorimetric assays, ion chromatography (IC), high-performance liquid chromatography (HPLC), liquid chromatography tandem mass spectrometry (LC-MS), enzymatic methods and electrochemical techniques (Kotani et al. 2023; Misiewicz et al. 2023; Varão Moura et al. 2022; Vasant Naik Bharati Vidyapeeth et al. 2014). Insoluble oxalates cannot be measured by IC (Cheng et al. 2016) and Raman spectroscopy is only suitable to quantify insoluble oxalates at high concentrations (Carmona, Bellanato, and Escolar 1997; Chiu et al. 2010; Khalil and Azooz 2007). Other techniques (HPLC, enzyme assays, electrochemical techniques) were not designed for clay mineral sample analysis, or the procedures are time-consuming, yield inconsistent results or require addition of Ca<sup>2+</sup> to promote precipitation (Cailleau, Braissant, and Verrecchia 2011b; Certini, Corti, and Ugolini 2000; Liu et al. 2015; Misiewicz et al. 2023; Mujinya et al. 2011; Shen et al. 2021; Vasant Naik Bharati Vidyapeeth et al. 2014). Liquid chromatography-mass spectroscopy (LC-MS) may be considered the preferred choice among the existing technologies due to high sensitivity and precision (Elgstoen et al. 2010; Keevil and Thornton 2006; Misiewicz et al. 2023), but to our knowledge the technique has not been used for oxalate analysis of geological samples.

Infrared (IR) spectroscopy provides a potential solution to the problems of oxalate quantitation (Aleixandre-Tudo et al. 2018; Gobrecht, Roger, and Bellon-Maurel 2014). IR analysis been shown to be more cost-effective compared to other techniques (Jozanikohan and Abarghooei 2022) and minimizes sample preparation and chemical waste, especially when an attenuated total reflectance (ATR) accessory is used (Aleixandre-Tudo et al. 2018; Bushong, Norman, and Slaton 2015; Mohamed et al. 2018; Pavia et al. 2015). IR spectroscopy is suitable for the analysis of heterogeneous samples when combined with chemometric techniques such as partial least-squares regression (PLSR) (Junaedi, Lestari, and Muchtaridi 2021b; Mohamed et al. 2018; Viscarra Rossel et al. 2006). Wavenumbers of absorption peaks commonly associated with biominerals, which are mainly detected in the mid-IR (MIR: 2500–25 000 nm) region, are shown in Table 1. IR spectroscopy has been used for semi-quantitative (Petrova et al. 2019) and quantitative analysis (Lin et al. 2019; Möller, Gergeleit, and Schüller 1989; Oka, Koide, and Sonoda 1985; Volmer, De Vries, and Goldschmidt 2001) of calcium oxalate in kidney stones, but no IR models have been developed for accurate quantification of oxalate salts in clay minerals or oxalic acid in solution. The aim of this study is to develop an efficient method to quantify CaOx and NaOx in a soil clay mineral matrix and oxalic acid in solution by calibrating

Wavenumber	Functional		
(cm <sup>-1</sup> )	group	Component	Reference
418	metal-O	NaOx	(Parekh et al. 2008)
515	C = O plane bend/Ca-O	Oxalate	(Bhatt and Paul 2008; Kachkoul et al. 2020)
610	metal-O	COD, NaOx	(Lin et al. 2019; Liu et al. 2020; Parekh et al. 2008)
650–670	C = 0	COM	(Kachkoul et al. 2020; Lin et al. 2019; Liu et al. 2020; Modlin and Davies 1981)
750–780	C = O out-of- plane bend	СОМ	(Bhatt and Paul 2008; Chen, Sun, and Zhou 2013; Kachkoul et al. 2020; La Russa et al. 2009; Liu et al. 2020; Modlin and Davies 1981; Rojas-Molina et al. 2015; Sikka et al. 2008)
885	C-C	COM	(Lin et al. 2019; Liu et al. 2020; Modlin and Davies 1981; Parekh et al. 2008)
912		COD	(Lin et al. 2019; Liu et al. 2020)
947–960		COM	(Liu et al. 2020; Modlin and Davies 1981)
1000–1800	C-C	COM	(Lin et al. 2019; Parekh et al. 2008; Schmitt et al. 2018)
1250–1425	C-C	COM	(Parekh et al. 2008; Sikka et al. 2008)
1310–1330	C = O symmetric stretch	Oxalate	(Bhatt and Paul 2008; Chen, Sun, and Zhou 2013; Kachkoul et al. 2020; La Russa et al. 2009; Lin et al. 2019; Liu et al. 2020; Modlin and Davies 1981; Rojas-Molina et al. 2015; Schmitt et al. 2018)
1383	C = O symmetric stretch	Oxalate	(Kachkoul et al. 2020; Sikka et al. 2008)
1590–1620	C = O asymmetric stretch	СОМ	(Bhatt and Paul 2008; Echigo et al. 2005; Kachkoul et al. 2020; La Russa et al. 2009; Lin et al. 2019; Liu et al. 2020; Modlin and Davies 1981; Parekh et al. 2008; Rojas-Molina et al. 2015; Schmitt et al. 2018; Sikka et al. 2008)
1646	C = 0	COD	(Kachkoul et al. 2020; Lin et al. 2019; Liu et al. 2020; Oyebiyi et al. 2018; Schmitt et al. 2018)
1660	C = O asymmetric stretch	Oxalate	(Bhatt and Paul 2008; Kachkoul et al. 2020; La Russa et al. 2009; Schmitt et al. 2018)
3286	O-H stretch (crystal H <sub>2</sub> O)	COD	(Echigo et al. 2005; Kachkoul et al. 2020)
3400-3500	O-H stretch (crystal H <sub>2</sub> O)	CaOx	(Bhatt and Paul 2008; Echigo et al. 2005; Kachkoul et al. 2020; Liu et al. 2020; Modlin and Davies 1981; Parekh et al. 2008; Schmitt et al. 2018)

Table 1. Wavenumbers of IR absorption peaks associated with biominerals CaOx monohydrate (COM), CaOx dihydrate (COD) and NaOx.

models with chemical and spectral reference data of standards to predict concentrations of oxalate components by ATR-MIR analysis.

# Materials and methods

An overview of the methods is presented in Figure 1. The degree to which the samples for calibration of the MIR predictive model is representative of future datasets is of greater importance than the size of the chemical reference dataset (van Wyngaard et al. 2021). Our models were calibrated using a small but targeted reference dataset.

## Sample collection and preparation

Clay mineral standards were prepared as mixtures (n = 34) consisting of minerals typically found in soils from a semi-arid region of South Africa where oxalate-rich vegetation occurs (Clarke et al. 2022; Francis 2008; Francis et al. 2013, 2015, 2020; Vermonti 2022). The standards consisted of soil clay minerals and oxalate salts at various concentrations ( $0-1.0 \text{ g g}^{-1}$  for CaOx and NaOx respectively) to test whether target analytes may be quantified reliably relative to other typical soil matrix components. The clay mineral mixtures were prepared by homogenizing milled analyticalgrade calcium oxalate monohydrate (CaC<sub>2</sub>O<sub>4</sub>·H<sub>2</sub>O, Thermo Scientific), sodium oxalate (Na<sub>2</sub>C<sub>2</sub>O<sub>4</sub>, Sigma-Aldrich), calcium carbonate (CaCO<sub>3</sub>, Sigma-Aldrich), gypsum (CaSO<sub>4</sub>·2 H<sub>2</sub>O, Kimix Chemical and Lab Supplies), quartz (SiO<sub>2</sub>, Sigma-Aldrich) and kaolinite (KGa-1b from the Source Clay repository at Purdue University, USA) powders using a mortar and pestle. The



Figure 1. Flow chart showing overview of the development of a mid-infrared (MIR) spectroscopic technique to predict oxalate concentrations in new samples using only spectral information. The model is calibrated by correlation between oxalate concentrations and spectral data of a reference dataset. Several models may be developed due to different possible combinations of predictive model components; thus, the best-performing model is selected by validation procedures.

quantities of the components in each standard are given in Table 2. Oxalic acid standards of various concentrations (0–0.01 M; n = 32) were prepared by dissolving analytical grade oxalic acid dihydrate (Merck) in water. The quantities of components in standard solutions are given in Table 2.

# Infrared spectroscopic modelling

#### IR model calibration

Spectral measurements of all solid and liquid samples were taken with an Alpha spectrometer (Bruker Optik, Ettlingen, Germany) operating in absorbance mode. A small quantity of each sample (~1 mL or equivalent mass of solid) was placed directly onto the ATR crystal surface of the MIR instrument. For solid samples, the probe was adjusted to make firm contact with the sample, while liquid samples were injected with a syringe via a sealed compartment. Sixty-four scans were collected in the range 350–7500 cm<sup>-1</sup> at a resolution of 4 cm<sup>-1</sup>.

MIR spectral-based models were calibrated using reference data derived from known theoretical concentrations of prepared oxalic acid solutions and clay mineral standards. Triplicate spectral scans of two standards (one clay mineral mixture as well as an oxalic acid solution) were included to incorporate spectral variability due to scattering of IR radiation by sample matrices (Aleixandre-Tudo, Nieuwoudt, and du Toit 2019; Nel, Clarke, and Hardie 2023). Multivariate analyses were conducted using OPUS 8.2 (Independent JPEG Group 2018). Models predicting oxalate components based on sample spectral measurements were calibrated by correlating IR spectroscopic data to corresponding reference data using a PLSR algorithm (Mohamed et al. 2018; Nel, Clarke, and Hardie 2023) with the QUANT2 function in OPUS 8.2 (Independent JPEG Group 2018) software. The oxalate components predicted by the models were CaOx and NaOx in solid clay samples and oxalic acid in standard solutions.

able 2. Composition of standard minera	al mixtures and oxalic acid solutions
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Mineral (g $g^{-1}$ )						Solution (M)
CaOx	NaOx					
$[CaC_2O_4 \cdot H_2O]$	$[Na_2C_2O_4]$	CaCO <sub>3</sub>	CaSO <sub>4</sub>	Kaolinite	SiO <sub>2</sub>	Oxalic acid
0.000	0.615	0.000	0.000	0.385	0.000	0
0.027	0.580	0.000	0.002	0.370	0.020	0.0003
0.037	0.523	0.044	0.002	0.333	0.061	0.0006
0.051	0.494	0.070	0.006	0.315	0.064	0.0009
0.065	0.455	0.087	0.006	0.295	0.093	0.001
0.084	0.424	0.115	0.006	0.269	0.102	0.0013
0.097	0.395	0.126	0.007	0.252	0.122	0.0016
0.109	0.367	0.145	0.008	0.236	0.134	0.002
0.123	0.347	0.148	0.008	0.213	0.160	0.0023
0.132	0.324	0.160	0.009	0.209	0.166	0.0026
0.142	0.299	0.178	0.008	0.191	0.181	0.003
0.153	0.282	0.182	0.011	0.170	0.202	0.0033
0.161	0.259	0.195	0.012	0.166	0.207	0.0036
0.164	0.231	0.221	0.013	0.153	0.217	0.004
0.180	0.212	0.232	0.014	0.135	0.226	0.0043
0.182	0.200	0.241	0.014	0.127	0.237	0.0046
0.195	0.178	0.246	0.015	0.115	0.252	0.005
0.196	0.168	0.252	0.016	0.109	0.259	0.0053
0.207	0.150	0.262	0.016	0.098	0.268	0.0056
0.218	0.138	0.267	0.016	0.090	0.271	0.006
0.224	0.125	0.277	0.017	0.077	0.280	0.0063
0.226	0.109	0.288	0.017	0.065	0.295	0.0066
0.235	0.095	0.294	0.018	0.060	0.297	0.007
0.236	0.087	0.302	0.019	0.051	0.304	0.0073
0.243	0.071	0.311	0.019	0.045	0.311	0.0076
0.246	0.060	0.314	0.020	0.034	0.325	0.008
0.254	0.041	0.321	0.020	0.031	0.332	0.0083
0.253	0.038	0.329	0.019	0.028	0.332	0.0086
0.259	0.031	0.337	0.019	0.018	0.335	0.009
0.265	0.018	0.345	0.020	0.010	0.342	0.0093
0.270	0.013	0.350	0.021	0.006	0.341	0.0096
0.272	0.000	0.355	0.021	0.000	0.352	0.01
0.531	0.000	0.000	0.000	0.469	0.000	
1.000	0.000	0.000	0.000	0.000	0.000	
0.000	1.000	0.000	0.000	0.000	0.000	
0.000	0.000	0.000	0.000	0.000	1.000	
0.000	0.000	0.000	0.000	1.000	0.000	
0.000	0.000	0.000	1.000	0.000	0.000	
0.000	0.000	1.000	0.000	0.000	0.000	
0.208	0.201	0.000	0.000	0.000	0.592	

The optimization procedure selected models with a combination of spectral pre-processing options with the lowest root mean square error of prediction (RMSEP)-values and a rank value <10 to ensure model parsimony (Pasquini 2018). The pre-processing techniques correct for baseline drift and light scattering (Lin et al. 2017; Nel, Clarke, and Hardie 2023). The best-performing model for prediction of each component of interest in the sample matrices analyzed in this study was reported in addition to figures of merit.

# **IR model validation**

Validation was performed using the QUANT2 function in OPUS 8.2 (Independent JPEG Group 2018). Calibration and test set samples were selected using the Kennard-Stone algorithm, which allocates samples according to desired proportions such that similar distributions of the property of interest are obtained in sample calibration and validation subsets (Ramirez-Lopez and Stevens 2020). Outliers were automatically detected according to the Mahalanobis distance (Kotu and Deshpande 2019) and metrics of models with lower RMSEP after exclusion of outliers were

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reported (<10% of original dataset). The ratio of performance to inter-quartile distance (RPIQ)value of each model in this report was calculated using the yardstick package (Kuhn and Vaughan 2020) in the RStudio 1.2.5033 computing environment (RStudio Team 2019). The RPIQ-value is an IR model performance metric like the ratio of performance to deviation (RPD) metric, but accounts for log-normal distribution of data often encountered in biological datasets (Johnson et al. 2019). RPD is calculated as follows:

$$RPD = \frac{SD}{RMSE}$$
 1

where SD = standard deviation of the calibration/validation sample set (for  $RPD_{cal}/RPD_{val}$  respectively)

RMSE = RMSECV/RMSEP (for RPD<sub>cal</sub>/RPD<sub>val</sub> respectively) (Aleixandre-Tudo et al. 2018)

RPIQ is calculated similarly to RPD, by replacing SD with inter-quartile distance (Bellon-Maurel et al. 2010).

The Quant2 Analysis function in OPUS 8.2 (Independent JPEG Group 2018) software was used to assess the reproducibility of spectral measurements of solid and liquid samples. This was achieved by applying the developed models for prediction of CaOx and NaOx content ( $gg^{-1}$ ) of a prepared mineral standard scanned in triplicate with the MIR instrument (repeat analyses of single sample) and calculating the degree of variability between replicates expressed as the coefficient of variation percentage (CV%). The models were also applied to predict oxalic acid concentration (M) of one oxalic acid standard scanned in triplicate with the IR spectrometer.

## Results

# Infrared spectroscopic analyses

The CV% values of MIR model-predicted oxalate salt components of the clay mineral standard (repeated scans of individual sample) were 1.1% and 0.5% for CaOx and NaOx, respectively. The CV% values of predicted oxalic acid concentration in the standard solution were 5.7%. Raw MIR spectra of an oxalic acid solution standard is shown in Figure 2. The absorption peaks



Figure 2. Unprocessed mid-infrared spectra of 0.01 M oxalic acid.



Figure 3. Unprocessed mid-infrared spectra pure solid minerals: calcium oxalate (CaOx, a), sodium oxalate (NaOx, b), kaolinite (c), SiO<sub>2</sub> (d), CaCO<sub>3</sub> (e) and CaSO<sub>4</sub> (f).

correspond to absorption of infrared radiation by hydroxyl (O-H,  $3500-3700 \text{ cm}^{-1}$ ) and carboxyl (C=O,  $1680-1720 \text{ cm}^{-1}$ ) functional groups (see Table 1). The MIR spectra of pure individual minerals used as components of clay standard mixtures are shown in Figure 3. The spectra show well-defined peaks corresponding to absorptions by carboxyl (655, 775, 1317, 1323, 1601 and  $1630 \text{ cm}^{-1}$ ) and metal-O (517 and  $528 \text{ cm}^{-1}$ ) functional groups of oxalate compounds (see Table 1). Crystal water O-H (3048, 3690 and  $3401 \text{ cm}^{-1}$ ) groups may be associated with hydrated metal oxalate compounds (see Table 1), kaolinite (Saikia and Parthasarathy 2010) and gypsum (Dabbas, Eisa, and Kadhim 2013). The Si-O stretching (1057, 1003 and 773 cm<sup>-1</sup>) and bending (449 and 411 cm<sup>-1</sup>) modes are visible in the pure quartz as well as kaolinite spectra (Tran et al. 2013). The peak at  $531 \text{ cm}^{-1}$  in pure kaolinite may be attributed to the Fe-O or Si-O-Al stretching mode, and absorption at 915 cm<sup>-1</sup> corresponds with OH deformation linked to Al (Saikia and Parthasarathy 2010). The pure gypsum spectrum shows S-O stretching

									Wavelength			
									regions	Pre-		Cal:
Component	Units	Matrix	R <sup>2</sup>	RMSEP	RMSEP <sub>norm</sub>	RPD	rpiq	Rank	(cm <sup>-1</sup> )	processing <sup>i</sup>	n"	val <sup>iii</sup>
CaOx	mmol	Mineral	0.98	0.217	0.02	7.38	5.69	7	7497.9–6781	MSC	42	2: 1
	$g^{-1}$	standard							3209.3-			
									2492.4			
									1779.7-			
									1062.9			
NaOx	mmol	Mineral	0.99	0.835	0.04	9.05	13.4	5	6068.3-3922	SNV	42	2: 1
	$g^{-1}$	standard							3209.3-			
									1777.7			
Oxalic acid	М	H <sub>2</sub> O	0.96	0.001	0.06	2.67	7.47	3	1779.7-	None	34	1: 2
		standard							1062.9			

Table 3. Figures of merit for models predicting oxalate components in solid and liquid matrices based on MIR spectra of samples.

<sup>i</sup>Spectral data pre-processing methods include first and second derivative (1<sup>st</sup> and 2<sup>nd</sup> der.), constant offset elimination (COE), straight line subtraction (SLS), vector normalization (SNV) and multiplicative-scattering correction (MSC).

n = Total number of samples for calibration and validation, including triplicate values of a single random sample for each of the prepared mineral mixtures and oxalic acid solution standards.

"Cal: val = calibration: validation sample splitting. A Kennard-Stone algorithm was used to split calibration and validation sample subsets such that both subsets included similar distributions of the measured component.

 $(1118 \text{ cm}^{-1})$  and bending  $(603 \text{ cm}^{-1})$  modes of the sulfate ion, while pure carbonate exhibits the stretching  $(1402 \text{ cm}^{-1})$  and bending  $(868 \text{ cm}^{-1})$  modes of the carbonyl functional group (Dabbas, Eisa, and Kadhim 2013).

#### Predictive model performance

Figures of merit for models predicting oxalate components in different matrices based on MIR spectra of samples are shown in Table 3. Validation graphs of these models for each oxalate component are illustrated in Figures 4 and 5. A very strong correlation exists between values predicted by spectral components and the true values of the measured property for all models ( $R^2 = 0.96$ , 0.98 and 0.99 for oxalic acid in solution, NaOx and CaOx in clay mineral mixtures). The RPD-values of models were larger than 2.0 (Table 3). Most RPIQ-values are higher than corresponding RPD values.

There are no absolute values of RMSEP that may be used as recommended thresholds of quantitative model accuracy, as RMSEP is reported in the same units as the target component measured. It is reasonable to assume that RMSEP smaller than 10% of the range of the component is indicative of relatively good predictive accuracy (Alexander, Tropsha, and Winkler 2015). Therefore, we judge models with RMSEP/range <0.1 to be suitably accurate for quantitative analysis of the measured property in this study. We define the RMSEP<sub>norm</sub> as the normalized RMSP, which is equivalent to the RMSEP/range value. Models for prediction of CaOx as well as NaOx in standard mineral mixtures have RMSEP<sub>norm</sub> values <0.1 (Table 3). Models predicting oxalate anions in standard solutions have RMSEP<sub>norm</sub> values <0.1 (see Table 3).

## Discussion

We observed low CV% (<5%) of CaOx and NaOx predictions in replicate IR measurements of the solid standard, which confirms excellent reproducibility of oxalate salts in a clay mineral mixture (Gomez and Gomez 1984). Relatively low CV% (<6%) in replicate IR measurements of the oxalic acid standard shows good reproducibility of spectral measurements of a liquid sample. The better reproducibility of MIR model predictions is attributed to spectral pre-processing procedures, as well as homogeneity of milled mineral samples, which reduces additive scatter of radiation by solid particulate solids (Naes et al. 2002). Excellent correlation between true and predicted CaOx and NaOx concentrations in compound clay mineral standards matches is equivalent to that obtained by IR analysis of CaOx in mineral mixtures with composition like that of human kidney stones



Figure 4. Plots for validation of infrared spectral-based models to determine concentrations of (a) calcium oxalate, CaOx and (b) sodium oxalate, NaOx in prepared mineral mixtures. Solid line represents y = x reference line.

 $(R^2 \ge 0.98)$  (Volmer et al. 1993). The IR models show slightly better fit than models that estimate the ratio of CaOx monohydrate to dihydrate in artificial mineral mixtures (Lin et al. 2019; Oka, Koide, and Sonoda 1985).

Wavelength regions selected by MIR spectral-based models to predict oxalate components in solid and liquid samples (Table 3) were mainly representative of carboxylic acid and crystal hydration functional groups. The "noise" observed closer to the further ends of the spectrum (Figures 2 and 3) contributes less meaningful information to models, which explains preclusion of these regions as predictive components by the PLSR algorithm. The carboxyl absorption at 655 cm<sup>-1</sup> and crystal water absorptions between 3000 and 3500 cm<sup>-1</sup> (Figure 3) is unique to the pure CaOx spectrum, creating a visible distinction between the metal oxalate salts. The 660 cm<sup>-1</sup> peak is



Figure 5. Plots for validation of infrared spectral-based models to determine concentrations of oxalic acid in a standard solution. Solid line represents y = x reference line.

a characteristic feature of CaOx that can be used to distinguish the monohydrate from the dihydrate form (Lin et al. 2019; Oka, Koide, and Sonoda 1985). Possibilities of overlapping peaks in mixed samples include the S-O bending mode of gypsum and carboxyl absorption of CaOx ( $603 \text{ cm}^{-1}$  and  $655 \text{ cm}^{-1}$ ), crystal hydration waters of CaOx monohydrate, kaolinite, and gypsum (3048, 3690 and  $3401 \text{ cm}^{-1}$ ), and Si-O stretching of silica and carboxyl of carboxyl absorptions by oxalates ( $773 \text{ cm}^{-1}$  and  $775 \text{ cm}^{-1}$ ). A slight overlap may also exist between the carbonyl group of carbonate ( $1402 \text{ cm}^{-1}$ ) and the carboxyl absorptions of oxalates ( $1300-1600 \text{ cm}^{-1}$ , Figure 3). Differences between wavelength regions used as predictive components by models for quantification of CaOx and NaOx (Table 3) show the capacity of the models to distinguish between the target analytes in compound materials based on absorptions in specific IR wavelength regions. This confirms the specificity of the IR technique for analyses of both CaOx and watersoluble oxalates in soil clay mineral mixtures.

The RPD and RPIQ of all models were greater than 2.0 (Table 3) which indicates reliable model performance (Bellon-Maurel et al. 2010). Low RMSEP<sub>norm</sub> values of models (RMSEP<sub>norm</sub> <0.05 for prediction of CaOx and NaOx; Table 3) confirm the accuracy of MIR spectroscopy to quantify oxalate salts in clay mineral mixtures. This level of accuracy was also achieved for IR analysis of CaOx in artificial mineral mixtures that mimic composition of human kidney stones (Volmer et al. 1993). Oxalic acid concentrations in solution were also predicted with good accuracy (RMSEP<sub>norm</sub> = 0.06, Table 3). These results provide evidence of the potential for MIR spectroscopy to be applied in quantitative studies of oxalate dynamics in soils. This technique should be tested in future on mineral samples that contain organic matter, which are more complex than pure mineral mixtures.

# Conclusions

This work demonstrates that MIR spectroscopic analysis in conjunction with chemometric data processing can be used to differentiate between soluble and insoluble oxalates in heterogeneous inorganic materials. To our knowledge, this is the first report of MIR spectral-based models to determine soluble and sparingly soluble oxalate salt concentrations in soil clay mineral mixtures and oxalic acid in solution. Our MIR models reliably quantify both CaOx and NaOx concentrations in compound clay mineral standards and oxalic acid in aqueous solution. We suggest that MIR

spectroscopy is a more efficient and cost-effective alternative to chemical methods of total and watersoluble oxalate salt quantification in geological samples.

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